

AD-A154 798 APPLICATION OF ADAPTIVE NOISE CANCELLATION TO COAST
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OFFICE OF RESEARCH AND DEVELOPMENT B B PETERSON ET AL.
UNCLASSIFIED MAR 85 USCG-D-5-85 F/G 17/2

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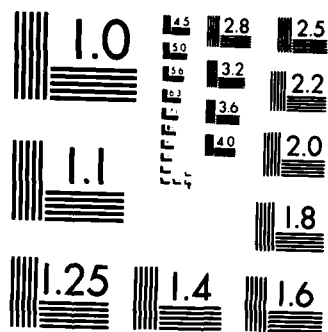
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Report No. **CG-D-5-85**

**Application of Adaptive Noise Cancellation
to Coast Guard Voice Communications**



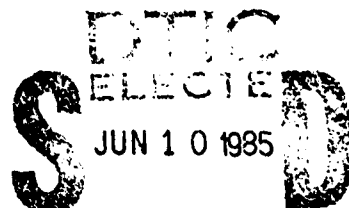
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16. Abstract <p>A variety of approaches to the digital filtering of voice signals corrupted by background engine noise are presented. These approaches include the standard Least Mean Square (LMS) adaptive noise cancellation algorithm, an optimum fixed weight filter, a special type of notch filter and frequency domain adaptive noise cancellation. The filters have been implemented both off-line using FORTRAN programs on an LSI-11/2 microcomputer, and in real time using the Texas Instruments TMS 320 microprocessor and in PDP-11 assembly language using an LSI-11/2.</p> <p>Frequency domain adaptive filtering was seen to be superior to the LMS time domain algorithm because its much greater computational efficiency allowed the analysis of much longer filters.</p> <p>A digital filter that exploits the engine noise by having a notch at each harmonic was seen to be more effective than any of the adaptive filters. The digital filter equations are derived, starting with a particular periodic reference input. This adaptive filter is shown to be, in reality, an infinite impulse response, time invariant filter.</p>			
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Abstract

A variety of approaches to the digital filtering of voice signals corrupted by background engine noise are presented. These approaches include the standard Least Mean Squares (LMS) adaptive noise cancellation algorithm, an optimum fixed weight filter, a special type of notch filter, and frequency domain adaptive noise cancellation. The filters have been implemented both off-line using FORTRAN programs on an LSI-11/2 microcomputer and in real time using the Texas Instruments TMS 320 microprocessor and in PDP-11 assembly language using an LSI-11/2.

Frequency domain adaptive filtering was seen to be superior to the LMS time domain algorithm because it's much greater computational efficiency allowed the analysis of much longer filters.

A digital filter that exploits the periodicity of the engine noise by having a notch at each harmonic was seen to be more effective than any of the adaptive filters. The digital filter equations are derived starting with an adaptive, finite impulse response filter with a particular periodic reference input. This adaptive filter is shown to be in reality an infinite impulse response, time invariant filter.

Table of Contents

I.	Introduction	1
II.	Experimental Methods	1
III.	Summary of Approaches and Results	3
	a. Analysis of LMS Adaptive Filters in FORTRAN	3
	b. Analysis of Optimum Fixed Weight Filters	4
	c. Real Time Adaptive Filtering Using the TMS 320	5
	d. Notch Filters	6
	e. Frequency Domain Adaptive Filtering	7
	f. Adaptive Filtering of Automobile Engine Noise	7
IV.	Publications and Presentations	11
V.	Conclusions and Recommendations for Future Research	12

Appendixes

Appendix A:	LMS Filters, Implementation and Performance	A1
Appendix B:	FORTRAN Program to Calculate Optimum Filter Weights	B1
Appendix C:	Implementation of an Adaptive Noise Cancellation Algorithm on the TMS 320	C1
Appendix D:	Notch Filter Analysis and Results	D1
Appendix E:	Frequency Domain Adaptive Filtering	E1
Appendix F:	LMS Adaptive Filter FORTRAN Program where Audio Signal is Added in Software	F1

References (Single list of references for both main body of report and appendixes)	R1
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I. Introduction

Background noise at the source has long been a problem affecting the intelligibility of Coast Guard voice communications. Recent studies of other researchers [7-10] indicate that additive noise will become a much more severe problem as the Coast Guard converts from analog systems to voice privacy systems using speech compression and digital transmission. In particular, Gold and Tierney at MIT's Lincoln Laboratory [7] have found that computer simulated F-15 cockpit noise added to clean speech reduces diagnostic rhyme test (DRT) [11] intelligibility scores to 92.6% without speech compression but to 75.2% when processed through a 2400 bit/second LPC-10 [12] speech compression algorithm.

Our efforts for the past eighteen months have focused on determining if digital filtering in general and adaptive noise cancellation in particular can improve the signal to noise ratio at the input to the system and thus the intelligibility of the received signal. A wide variety of approaches have been attempted with varying degrees of success. A summary of these efforts is contained in the main body of this report. The specific details including program listings are in Appendixes A-F.

II. Experimental Methods

The general technique employed has been to make stereo tape recordings of background engine noise both with and without a voice signal on one of the channels, and then to process and

and were made

in the laboratory.

Recordings were made

my Luder yawl with the engine (a four

el) running, in the Academy Power

with a single cylinder, four cycle,

running and most recently in the Power

with a four cylinder Datsun 710 running.

Reel to reel tape deck donated by CDR

for the recordings. All recordings were

General Radio Model 1952 Universal

pass anti-aliasing and reconstruction

essing the recorded waveforms were of two

gments (less than one second) were

emory using a Digital Equipment

rocomputer and a Data Translation 2785

Filters were then implemented and the

ie using FORTRAN programs.

as to implement the filter in real time

on an oscilloscope or to listen to the

one exception this was done using a

20 Evaluation Module (EVM). The notch

ndix D was implemented in real time on

the TMS 320 EVM. In our proposal [13]

real time implementation were proposed.

W TDC-1010 hardware multiplier/

ed to the LSI-11/2 computer but even with

slow to implement reasonable size

adaptive filters in real time. It was hoped to play the tape deck into the filter at a much slower speed than it was recorded, tape record the filter output and then play the output back at the original speed, but we were unable to obtain a tape deck with this capability. No attempt was made to use the Western Digital WP 3150 Programmable Digital Filter. We were unable to obtain a final data sheet and have concluded the announcement was a trial balloon and that the circuit never went into mass production. An overwhelming conclusion we have reached in the context of our work is that the TMS 320 is by far the best approach to audio frequency digital signal processing. The second generation of the circuit [14] was announced in February 1985 at the IEEE International Solid State Circuits Conference and will make even more sophisticated processing possible.

III. Summary of Approaches and Results

a. Analysis of LMS Adaptive Filters in FORTRAN

Our first approach was an off-line FORTRAN implementation of the standard LMS adaptive noise cancellation algorithm [1]. Several such programs were written, one of which is described in Appendix A. When the inputs were highly correlated signals from laboratory signal generators the filter was seen to be very effective. When the inputs were two microphone recordings of engine noise there was virtually no improvement in signal to noise ratio. It is felt this was due to a combination of two problems. The filter length had to be kept short to insure the weights would converge within the sample size allowed by the 56

Kbyte random access memory of the LSI-11/2. This resulted in filter lengths much shorter than the fundamental period of the engine noise and very poor filtering. When the filter length was increased the weights would not converge within the allowable sample size. The recent acquisition of an LSI-11/23 microcomputer with 248 Kbytes of random access memory from the USCG Electronics Engineering Laboratory will permit the analysis of longer filters in the future.

b. Analysis of Optimum Fixed Weight Filters

It was next attempted to eliminate the slow weight convergence problem by calculating the optimum (in a linear least squares sense) fixed weight filter for the particular data set and then calculating the output using these fixed weights. A description and listing of the FORTRAN program to accomplish this is contained in Appendix B. Improvements in signal to noise ratio of 6-8db were possible when the sampling rate was reduced to 1-2 kHz and the anti-aliasing filters adjusted accordingly. This effectively increases the filter length in time to approximately the period of the noise. Although these low sampling rates are unacceptable for voice communications the intent was to determine if longer filters would be effective at the necessary higher sampling frequencies. It is also believed there were numerical problems in the Gauss elimination subroutine when attempting to analyze longer filters. The subroutine should have used double precision arithmetic [15] but single precision was used due to limited memory and because double precision floating point instructions are not part of the assembly language

instruction set on the LSI-11/2 [18] and would have been slow to execute. The LSI-11/23 with it's larger memory and double precision instructions in assembly language will allow analysis of longer filters.

c. Real Time Adaptive Filtering Using the TMS 320

At this point (November 1983) a TMS 320 Evaluation Module was procured and efforts shifted to real time implementation. The first version was done by then Cadet 1/c now ENS M. D. SAKAHARA as his Systems Design and Synthesis project. His report was included in the July 1984 interim report [16]. His filter had 40 weights and would operate at sampling frequencies of up to 8.5 kHz.

A second version (Appendix C) used BASIC on the Dartmouth Time Sharing System to write much more efficient TMS 320 code and resulted in a 68 weight filter that would operate at sampling rates of up to 10.7 kHz. The essential difference is that all the looping is done in BASIC at the assembly language generation level resulting in straight line code that executes at the maximum benchmark speed of 1.2 microseconds/weight [17]. The first version used conditional branching and executed at 2.6 microseconds/weight.

Extensive tests were done on recorded engine noise and the results are contained in Appendix C. The conclusion reached was that 68 weights are not enough for an 8 kHz sampling frequency. The second generation TMS 32020 [14] has 544 words of on chip data memory compared to 144 for the TMS 32010, efficient access to 64K words of external memory, and has combined some operations

that were two instructions into a single 200 ns instruction. Adaptive filters of at least 120 weights at 8 kHz should be possible.

d. Notch Filters

The highly periodic nature of the engine noise implies that any reference input that contains all the harmonics of the noise in the primary input should suffice. A reference input of unit impulses periodic at the same frequency as the engine noise contains these harmonics and results in very efficient algorithm allowing filters of virtually arbitrary length. Synchronization is obtained by triggering the A/D converter with an optical shaft encoder mounted on the engine.

Further analysis of the algorithm revealed that what started as an adaptive, finite impulse filter was in fact a fixed weight, infinite impulse response filter. Preliminary results of using the filter were contained in the interim report [16] and a detailed analysis is contained in Appendix D.

This filter appears to be the most promising at this stage. It is intended to further evaluate it's performance by making a series of comparative diagnostic rhyme tests (DRT's) [11]. Comparisons of intelligibility will be made among filtered and unfiltered speech using both an omnidirectional microphone and a Shure Model 562 noise canceling microphone [19-20]. Due to the failure of the motor generator set and therefore the lack of DC power it has not been possible to start the engines in the Power Engineering Laboratory since September 1984. We have designed and constructed a high power DC supply and hope to have engines running in the near future.

The LMS filter was implemented through simulation on the LSI-11/2 computer. Since the LSI-11 could not execute the adaptive algorithm fast enough for real time applications, data sets were obtained and the adaptive algorithm applied to the data sets.

A Fortran IV program was written to implement the adaptive filter and used assembly language subroutines to communicate with the real world (see flow chart). The input data set was placed into two input arrays, one for the samples with noise (IN1) and the other for samples containing noise plus the desired signal (IN2).

The program first implements a fixed weight filter on the reference input and subtracts the result from the primary input. The output of the adaptive filter was plotted and displayed on the video terminal. This result was applied to the adaptive algorithm to adjust the weight vector. Once the adaptation cycle is completed, the computer then gets the next input and starts the process all over.

When all of the primary samples are used, the program plots the impulse response of the filter and calculates the frequency spectrum of the weights. Both phase and magnitude were plotted. Since the adaptive cycle was applied after the fixed filter, the plotted weight vector has gone through an extra adaptation cycle.

In the strictest case, the performance of the plotted weights is unknown. However, this does not pose a significant problem since the weights change very little once they have converged. This is evident in the plots obtained from the program (see Appendix 1). Once the weights converged on the LMS solution, they change very little.

The entry of pertinent constants is executed at the beginning of each run. On the first run, the input channels (IN1 and IN2) were selected and the number of samples (NUMS), delays (NUMD) and weights (NUMW) for the filter entered. The adaptive coefficient (ADAPT) was entered and the weight (WEIGHT) and input arrays were reset as necessary. The data entry segment of the program was placed at the end of the program to help speed execution, however it did not have a significant effect.

Implementation

The least mean square (LMS) algorithm of the adaptive filter was used since it was one of the more simple forms. It has the form of

$$W(k+1) = W(k) + u Y U(k)$$

where $W(k)$ is the weight vector, Y is the output of the adaptive filter, $U(k)$ is the reference input array and u is the adaptive coefficient.

The reference signal was filtered and then subtracted from the primary signal. The result is treated as an error signal for the filter. For a positive output, the adaptive algorithm will tend to increase the magnitude of the weight vector. The output of the filter will increase causing the output of the adaptive filter to decrease or tend negatively. When the output goes negative, the algorithm will again tend to alter the weights to decrease the output. This tendency to minimize the output is where the LMS algorithm gets its name.

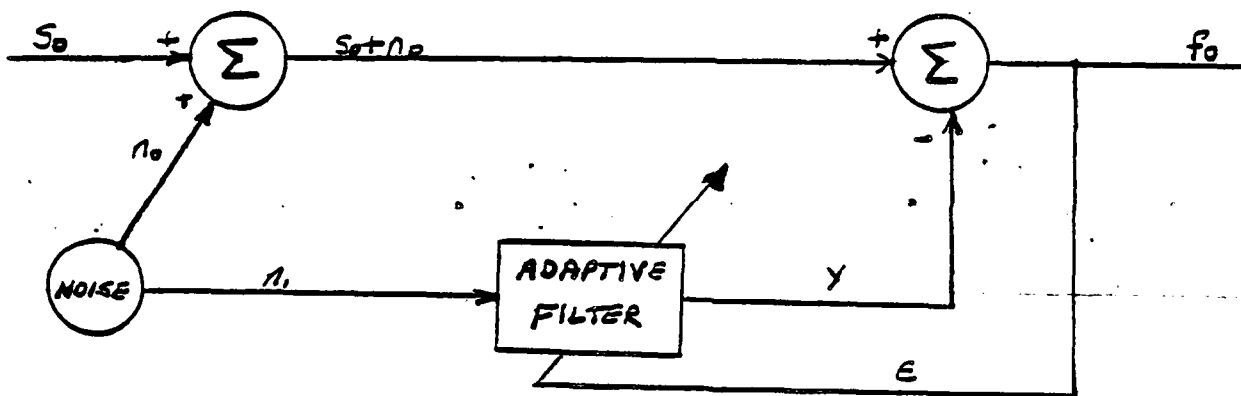
It would seem that if this filter were working in the ideal case, the weights would converge to completely cancel the output. The adaptive coefficient is what prevents this from happening. The magnitude of the adaptive coefficient must be small. Since the adaptive coefficient has a direct affect on the rate of convergence of the weights, a small value for u would reduce the tendency to completely cancel the output and yet suppress most of the interference.

The cancellation of the noise in the primary input is accomplished by reshaping the reference input, through filtering, to reflect the noise in the primary input. If the filter is to adjust its characteristic to accomplish this reshaping, it would be desirable for the characteristic to change slowly. This avoids oscillating around the desired response as the case would be if the weights were allowed to converge too quickly. A large adaptive coefficient may reduce the effectiveness of the overall filter to the point of uselessness (greater than $2.0 \cdot 10^{-9}$ in this application) or cause the filter to become unstable and blow-up (u greater than 2.0). A coefficient that is too small will impede the weights from converging or even completely preventing them from reducing any noise in the primary input. A compromise must be reached between rate of convergence and accuracy of the convergence in the proper selection of u . The adaptive coefficient must be small enough to allow the filter to accurately reshape the reference input and allow the weights to converge in a reasonable time.

Introduction

A common method of estimating a voice signal distorted by additive noise is to pass it through a filter that tends to reduce the noise level in the signal. This filter can either be fixed or adaptive. The fixed filter requires a prior knowledge of the signal and noise in order for it to be effective.

The adaptive filter does not require any or little knowledge of the signal or noise. It has the ability to alter its impulse response automatically to suppress the noise from the signal. The adaptive algorithm used in this application has two inputs: a primary input that contains the signal (s_i) and noise (n_i) and a reference input that contains noise (n_r) that is somehow correlated with the noise (n_i) in the primary signal. The reference input should contain none or very little of the signal since the filter may converge to cancel to signal as well as the noise in the primary input.



BLOCK DIAGRAM OF ADAPTIVE FILTER

The adaptive filter can be used to eliminate the 60 Hz hum in EKG's. The 117vac outlet can be used as the reference input and the normal chest leads as the primary input. In a similar application, the mother's heartbeat can be filtered away from an infant's EKG. The reference leads are placed near the mother's heart and the primary leads are placed on the mother's abdomen. Interference entering the side lobes of a receiving antenna can be suppressed by using a circular array of primary antennas and a reference antenna chosen in the area of the incoming interference. In voice communications, where background noise is a problem, two microphones may be used. The operator speaks into the primary microphone while the reference microphone is placed in the relative vicinity of the operator picking up the background noise.

Table of Contents

Abstract	2
Introduction	4
Implementation	5
Data and Results	7
Interface design of TRW's TDC1010J to the LSI-11	9
Flow chart of Fortran IV program	11
Appendix I	13

Abstract

An LMS adaptive filter was implemented on the LSI-11/2 computer. The computer could not process fast enough for a real time application so the filter was simulated on input data sets already contained in the computer's memory. The filter used two inputs: a primary input that contained noise and a desired signal and a reference input that contained noise that was somehow correlated with the noise in the primary input. The reference input was filtered and subtracted from the primary input suppressing the noise. The filter produced signal to noise improvements of around 20 dB in most cases and worked quite well with an adaptive gain of $5.0 \cdot 10^{-4}$.

LMS FILTERS IMPLEMENTATION AND PERFORMANCE

By Cadet 1/C Michael Sakahara

Instructor : LCDR Benjamin B. Peterson

Fall 1983

Department of Applied Sciences
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affect of background noise on the parameters internal to the algorithm and this is considered essential to basic understanding of the problem.

cancellation compared to noise canceling microphones but adaptive noise cancellation where the primary input is a noise canceling microphone vs a single noise canceling microphone. In related work for the Air Force researchers at Bolt, Beranek, and Newman, Inc. [22] have experimented with various multisensor configurations and have found a combination of a noise canceling microphone at the lips and an accelerometer attached to the throat to be best. The accelerometer is insensitive to environmental noise but is bandlimited to 2 kHz. The accelerometer output was added to the microphone output high pass filtered at 1200 Hz. Also, in work for the Navy, Ketron, Inc. [23] has developed a second order gradient noise canceling microphone (NC-104LF). Their tests have shown substantially improved performance over first order gradient microphones.

As noted in the introduction, background noise is a more serious problem in digital voice privacy systems employing the LPC-10 speech compression algorithm [12]. Therefore, future efforts should incorporate the LPC-10 algorithm and measure intelligibility after synthesis. Cadet 1/c P. A. MENDEL is presently attempting to implement LPC-10 in FORTRAN IV on an LSI-11/2. We have also requested the PDP-11 FORTRAN 77 program referred to in [12] from the National Security Agency and we hope to modify it for running on an LSI-11. These programs will not run in real time as required for DRT's. The best method of real time implementation would be a TMS 320 based coprocessor board for an LSI-11. An alternate method would be to acquire two digital voice privacy units but this would not allow access the

weight FIR and IIR filters, plot their impulse and frequency responses and automatically generate TMS 320 assembly language code for downloading and real time implementation. The program for FIR design is that described in [21] and installed on DTSS when PROF WOLCIN joined the Academy faculty in 1984. Cadet 1/c B. M. LAM is presently modifying the program to produce TMS 320 code for real time implementation.

V. Conclusions and Recommendations for Future Research

We have demonstrated that under some conditions digital filtering can improve the signal to noise ratio of speech corrupted by background engine noise. There are several questions suggested by our research that remain unanswered.

Does improved intelligibility result from this improved SNR? Diagnostic rhyme tests [11] are necessary to resolve this. DRT's are quite simple and straightforward but will require many man-hours of effort. Cadet projects may be the best method of performing DRT's. How intelligibility is measured is worth learning and the data analysis is a good exercise in undergraduate statistics.

The type of analysis done in the context of this project has not lent itself to a good comparison of adaptive noise cancellation vs noise canceling microphones. Again, this sort of question can only be answered using DRT's. The general consensus among experts in the field is that noise canceling microphones are essential in high ambient noise environments independent what else is done. Therefore the basic question is not adaptive noise

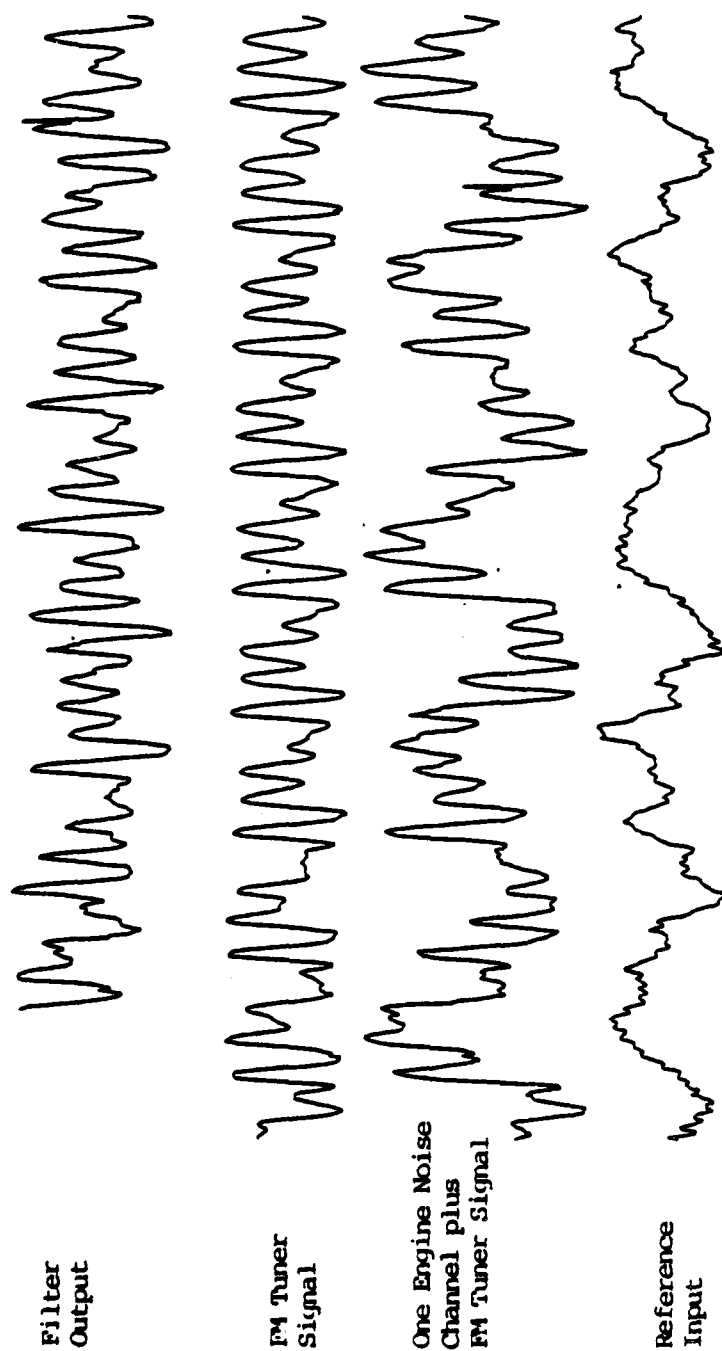
Conversations with other researchers in the field indicate that improved signal to noise ratio may not necessarily imply improved intelligibility and that DRT's are necessary to evaluate a proposed system. The human auditory system is very sophisticated and can extract information from highly corrupted signals but non-linear processing may destroy cues the human system is using not improve intelligibility.

IV. Publications and Presentations

Presentations of papers based the research have been made at two IEEE sponsored conferences and one local IEEE meeting to date and a fourth abstract has been submitted for presentation at a future conference. As noted in [16] M. D. SAKAHARA presented his paper "Adaptive Noise Cancellation" at Electro '84' at Boston in May 1984. He also presented "Real Time Adaptive Noise Cancellation" at the April 1984 IEEE New London Subsection meeting. This paper was included in [16].

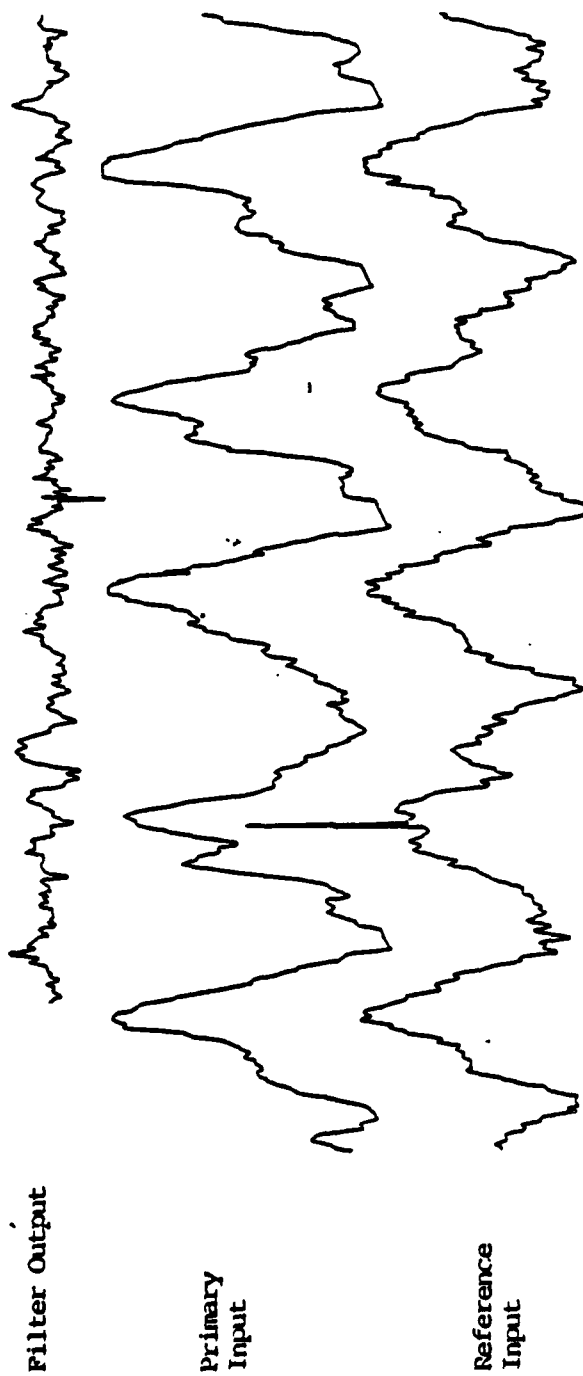
B. B. PETERSON presented "Audio Frequency Adaptive Filtering Using the TMS 320 Microprocessor" at the IEEE Digital Signal Processing Workshop at Chatham, MA in October 1984. In addition K. U. DYKSTRA and M. D. SAKAHARA were coauthors of the published summary and LCDR DYKSTRA attended.

An abstract titled "An Integrated Approach to the Design and Implementation of Digital Filters" has been submitted to the IEEE/ASEE Frontiers in Education Conference at Golden, CO in October 1985. In addition to the BASIC program in Appendix C, the paper will describe interactive programs to design fixed



Results of Adaptively Filtering Automobile Engine Noise when Audio Signal is added in Software

Figure 2



Results of Adaptively Filtering Engine Noise
No audio signal added

Figure 1

filtering the reference input with an 80 weight filter and subtracting. The sampling frequency was 8 kHz.

The program was then modified to sample a third channel which was the audio output of an FM tuner. In software this signal was then amplified and added to the primary input and resultant signals run through the adaptive filter algorithm. Appendix F is this FORTRAN program. A listing of the graphics package written to drive the DEC Gigi Graphics Terminal and the HP 7470A Plotter is not included because of it's length and because the subroutine calls are obvious. This package was the most extensive programming effort undertaken in the context of this research and the excellent work of LTJG W. M. DUPRIEST is acknowledged.

Unlike previous efforts, it is now possible to know exactly what is signal and what is noise. Figure 2 is a sample plot of the results. The signal to noise ratio has been improved from -3.3 db to +6.0 db. Improvements in signal to noise ratio were consistently near 10 db. Reductions in noise power of 20 db as noted when no signal was added (Figure 1) are not expected because what is signal to the overall filter algorithm is actually noise to the weight update algorithm. The fluctuation in weights caused by the larger output degrades the filter performance. This effect is minimized by reducing the adaptive gain at the expense slower convergence and poorer tracking in time varying conditions.

What these recent results do indicate is that adaptive filtering can work very well when the noise is highly correlated.

e. Frequency Domain Adaptive Filtering

Just as their fixed weight counterparts, frequency domain adaptive filters require only a fraction of the arithmetic operations required in a time domain filter of equivalent complexity. The frequency domain adaptive filter algorithm described in [5] was implemented in FORTRAN for filter lengths of up to 512 samples. Due to filter length and good convergence properties the filter performance was better than that of the adaptive filters in a. and c. above but not as good as the notch filter in d. Detailed analysis of the filter is contained in Appendix E.

DRT's using a real time version of the algorithm would be necessary to make a more complete evaluation. The TMS 320 is fast enough but due to limited memory, implementation with our existing hardware is not possible.

f. Adaptive Filtering of Automobile Engine Noise

Our most recent effort was to make a two microphone recording of engine noise from a Datsun 710 with no muffler running at idle (approximately 1000 rpm) in the Power Engineering Laboratory. Due to the faulty exhaust system, it was necessary to operate the ventilation system at full power adding another significant component of noise.

When the data was plotted in the laboratory, the two signals were seen to be much more correlated than had been observed in previous recordings. Figure 1 shows two signals and the results of adaptive filtering with no voice signal added. 98.9% of the power in the primary input has been canceled by adaptively

Results and Conclusions

Pure signals (sine, square and triangle) were used to simulate noise and desired signal. A square wave was used to simulate noise since it contained the most harmonics and could be shaped to any of the noise signals placed in the primary input.

Three Wavetek signal generators were used to supply the noise, signal, and clock pulses for the sampler. Hewlett-Packard frequency counters and oscilloscopes were used to monitor the inputs and outputs of the computer simulation. In the simulations, the noise in the reference and primary inputs were of the same frequency and phase. This simplifies predicting the impulse response the filter should converge to allowing for an evaluation of the filter's performance.

The primary input contained a square wave as the desired signal. This enables the calculation of signal to noise ratios (SNR) based on the amplitudes of the noise and the desired signal. The SNR improvement of the filter could be estimated and the performance of the filter evaluated based on the SNR improvement.

A triangle wave represented the noise in the primary input. This simplified predicting the impulse response the filter should converge to. The Fourier series of a square of unity amplitude is

$$4/\pi (\sin wt + 1/3 \sin 3wt + 1/5 \sin 5wt + . . .)$$

and a triangle wave of same amplitude

$$3/\pi (\sin wt - 1/9 \sin 3wt + 1/25 \sin 5wt - . . .).$$

In order to change the square wave in the reference input into the triangle wave in the primary input, the square wave must be multiplied by another square wave of the same frequency.

When the weights were plotted, this indeed seemed to be the result. There were small deviations in the impulse response of the filter and was evident in the incomplete cancellation of the noise in the primary input. Although noise cancellation was not total, SNR improvements of up to 32 dB were observed.

The adaptive coefficient was varied and had a proportional affect on the convergence time of the filter. When u was $1.25 \cdot 10^{-9}$, the filter converged in 12 msec. When u was $4.0 \cdot 10^{-7}$, the convergence time was 45 msec or it took about four times as long. In both cases, SNR improvements of 27 dB was observed.

When μ was increased beyond $2.7 \cdot 10^{-3}$, the filter seemed to have no effect on the primary input. Since the adaptive coefficient was so large, the weights constantly over shot the desired solution and did not reduce the noise levels in the primary input. The adaptive gain was not sufficiently high enough to cause instability, just large enough to cause excessive oscillations. When the adaptive gain was increased beyond 2.7, the filter quickly became unstable and blew-up. This was the threshold where the weights were allowed swing so far that they overcame the reducing tendencies of μ .

The advantages of an adaptive filter were especially evident in the case when the noise and signal were similar in frequency. A run with the noise at 500 Hz and the desired signal at 750 Hz produced a SNR improvement of 32 dB!

The LMS filter seemed to suffer when the noise was at a significantly lower frequency than the signal. The filter would "glitch" every time the triangle wave would change slope. The filter would quickly converge to the desired signal while the slope was constant, however would loose track when the slope changed. This is due to the fact that the LMS filter follows the gradient of the noise. A low frequency triangle wave has a low gradient while scribing constant slope, however the gradient greatly increases when the triangle wave changes slope.

Over all, the LMS algorithm performed quite well in this application. It produced SNR improvements of around 27dB and exceeded 30 dB in some cases. The adaptive coefficient was critical in the performance of the system. The lower frequencies tended to require higher adaptive gains to optimize the filter, while the higher frequencies required lower adaptive gains. The adaptive gains depended mostly on the contents of the reference input and the noise frequencies. The adaptive gains seemed to be independent of the desired signal. An adaptive gain of $5.7 \cdot 10^{-4}$ worked the best for all cases of noise and signal combinations. The LMS filter worked well for most combinations of noise and desired signals.

Design of interface between LSI-11/2 and TRW's TDC-1710J

A cookbook approach was used in designing the interface between TRW's TDC-1710J Multiplier Accumulator chip and the LSI-11/2 computer. MDS Systems general purpose interface module (MSI-1717) was used to mount the TDC-1710J and supporting hardware.

The interface module provided input, output and address ports that were common on the Q-bus of the LSI-11. This simplified the decoding and interface logic greatly since bus protocol did not have to be considered. It was determined that the TDC-1710J could multiply and accumulate within the time of a DATA cycle of the LSI-11. Thus the only timing considerations made were in delaying signals going into the TDC-1710J.

The Y multiplicand and the least significant 16 bits of the output share pins on the TDC-1710J. Logic had to be provided to isolate the input and output ports on the interface module. Noninverting tri-state bus transceivers (Ti's 74LS243) were used to isolate the ports. The B side of the transceivers were connected to the TDC-1710J and the A side to the respective I/O location. The control inputs for the transceivers are Gab (L) and Gba (H). When Gab is asserted, the data is allowed to flow from the A side to the B side. When Gba is asserted, the data is allowed to flow from the B side to the A side. Asserting both control inputs could result in destructive oscillations because the transceiver is trying to pass data in both directions at the same time. Not asserting both control inputs isolates side A from side B.

In the case of a Y input, the Gba input is held low not asserting it and Gab was asserted when the input data was valid. The reason Gba was held low was to prevent the possibility of sending the transceiver into destructive oscillations. By keeping Gba low, the transceiver can be alternated from isolation to passing data from A to B (the desired direction). The logic for the Gab input was (BDOUT * ADD3) (L). ADD3 is the address location indicating a Y-input on the bus and BDOUT is asserted when the data in the input port is valid. The Gab signal was used to supply the CLK Y signal. The three inverters delay the CLK Y signal to account for propagation delay of the transceivers.

Since the X-input shares the input port with the Y-input and control byte, it was isolated from the input port when not in use. This was accomplished by using tri-state noninverting bus drivers (Ti's 74367). This circuitry is not absolutely necessary because the X-input is loaded on to the TDC-1710J only when CLK X is asserted. But was added as a precautionary measure

only. The control logic for the bus drivers is (ADD2 • BDOUT) (L). ADD 2 is the address location indicating an X-input will appear on the input port and BDOUT is asserted when the data on the input port is valid. This control signal was used to supply the CLK X signal with inverters to delay to signal to account for propagation delays in the bus drivers.

The TDC-1010J has five data control bits:

Accumulate	(ACC)	bit 00
Subtract	(SUB)	bit 01
Preload	(PREL)	bit 02
Two's Complement	(TC)	bit 03
Round off	(RND)	bit 04

all asserted high. It was decided for versatility that these control bits be controlled by the LSI-11. The control bits are loaded onto the TDC-1010J at different times making it difficult or impossible to send them with the data going into the board. Type D flip flops (Ti's 7474) are used to store the control byte. The clock signal for the flip flops is (ADD1 • BDOUT) (H). ADD1 is the address location indicating that the control byte is to appear on the input port and BDOUT is asserted when the data on the input port is valid.

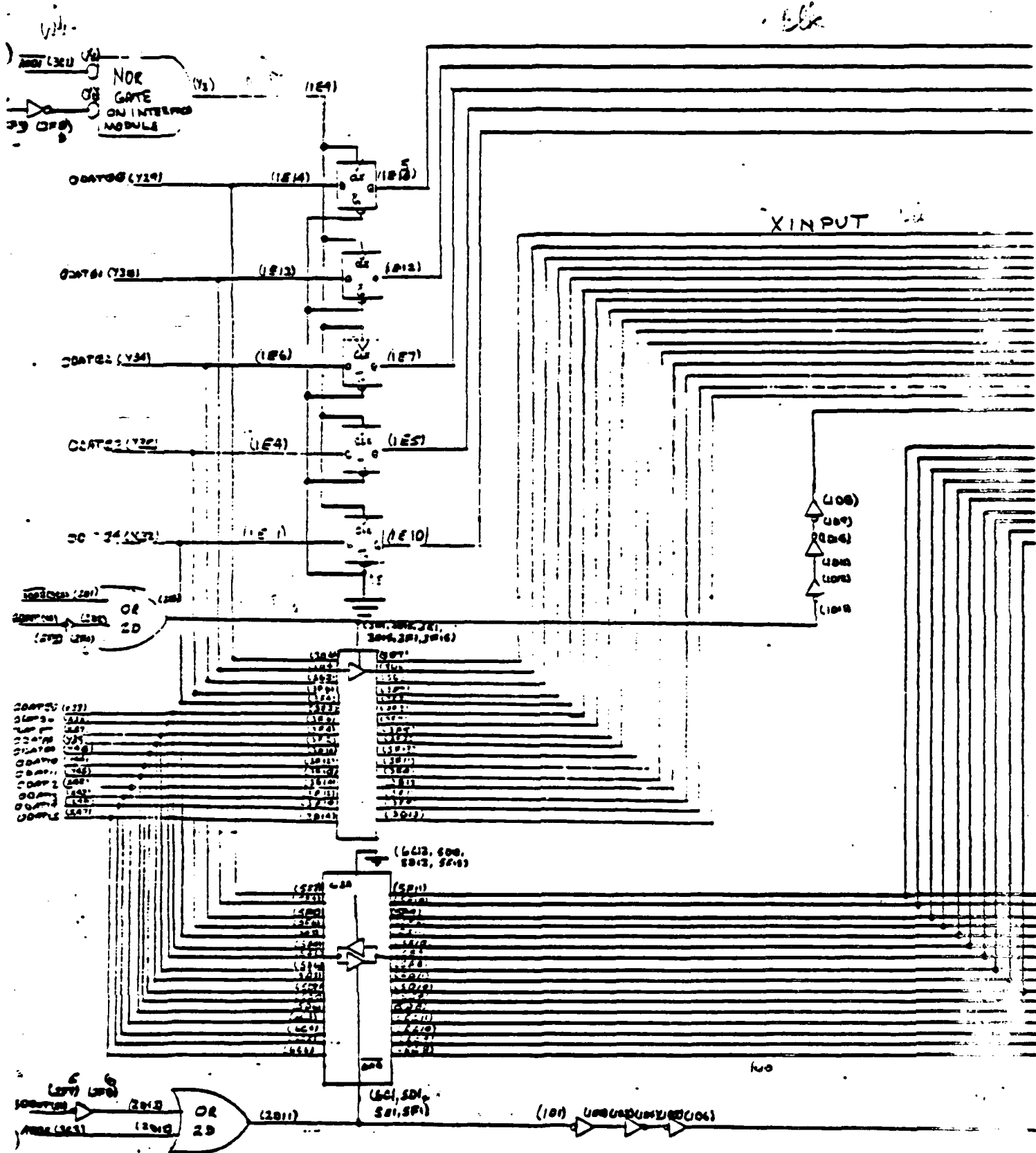
The output of the TDC-1010J is divided into three 16 bit words: the least significant product (LSP), most significant product (MSP) and extended control product (XTP). The LSP is time shared with the Y-input and similar circuitry is used to isolate the input and output ports as in the Y-input. The tri-state output of the TDC-1010J is controlled by three inputs TSX, TSM and TSL. PREL must be held low if the TDC-1010J is to generate output. The control inputs are asserted low and only one of them may be asserted at a time for this design. A separate address has been allocated for each output word with TSM clocking the output to the output registers. The output of the address decoder on the interface module is asserted low and is connected directly to the appropriate output control pins. The interface module requires an input enable signal to indicate that the data on the output port is valid. This is accomplished by ORing the output control signals and using an inverter to account for propagation delays.

Six address locations were used to implement the TDC-1010J :

Address	Function number	Location
0	CTRL Byte	170000
1	X- Input	170010
2	Y- Input	170002
3	LSP Prod.	170012
4	MSP Prod.	170004
5	XTP Prod.	170014

The unusual order of address locations is due to the address decoder on the interface module. The decoder an AND gate for 3 significant bits and a BCD to Decimal (Ti's 7442) for the lower three bits. Bits 01 and 02 are hard wired to the B and C inputs and the output of the AND gate is wired to the D input of the decoder. Bit 03 was wired to the A input and resulted in the unusual address order. Bit 03 was brought to the decoder to enable it to decode eight address instead of four.

To use the TDC-1010J , the output registers must be zeroed. This can be accomplished by sending a 0 to the X-input and multiplying without accumulation. The MSP must be outputted to a dummy location to clock the output registers. The other functions of the TDC-1010J can then be executed by altering the control byte as necessary. When ever output, accumulation or subtraction is desired, the MSP must be moved to a dummy location since asserting the MSP clocks the output clock and activates the accumulator and clocks the output registers.





```
C PROGRAMMER: Michael D. Sakahara
C
C DATE: November 13, 1983
C
0001 PROGRAM ADAPT
C THIS PROGRAM IMPLEMENTS AN ADAPTIVE FILTER ALGORITHM. THE FILTER
C IS IMPLEMENTED IN THE BEGINING OF THE PROGRAM AND THE USER INPUT
C PORTION OF THE PROGRAM APPEARS AT THE END OF THIS PROGRAM.
C
C
C INSTRUCTIONS: WHEN ANSWERING YES TO ANY OF THE QUESTIONS, ENTER
C ANY VALID INTEGER, HOWEVER WHEN ENTERING NO ENTER A '1' ONLY.
C
C
0002 DIMENSION WEIGHT(100), IN1(5000), IN2(5000), REAL(201), PHA(201)
0003 GOTO 50
C
C ADD DELAY TO AVOID INTERRUPT FROM KEYBOARD
C
0004 10 DO 100 I=1, 10000
0005 100 CONTINUE
0006 DO 110 I=1,5000
0007 IN1(I)=0
0008 IN2(I)=0
0009 110 CONTINUE
C
C GET NUMS SAMPLES FROM A/D CONVERTER
C
0010 CALL XTSMF2 (NUMS,NC1,NC2,IN1,IN2)
C
C ADD DELAY TO AVOID INTERRUPT FROM KEYBOARD
C
0011 20 DO 120 I=1, 10000
0012 120 CONTINUE
0013 DO 130 I= NUMW+1,NUMS-NUMD
0014 SUM=0.0
C
C IMPLEMENT FIXED WEIGHT FILTER
C
0015 DO 140 J=1,NUMW
0016 SUM=SUM+WEIGHT(J)*IN1(I-J)
0017 140 CONTINUE
0018 K=IN2(I+NUMD)
0019 OUT=K-SUM
0020 WRITE (7,800) K,OUT
0021 800 FORMAT (1X,I8,G12.5)
C
C OUTPUT OF FIXED WEIGHT FILTER TO D/A (INPUT, OUTPUT)
C
0022 CALL DAOUT(K,INT(OUT))
C
C IMPLEMENT ADAPTIVE FILTER
```

```

0023      W=ADAPT*OUT
0024      DO 150 J=1, NUMW
0025          WEIGHT (J)= WEIGHT (J) +(W*IN1(I-J))
0026 150    CONTINUE
0027 130    CONTINUE
0028      CALL DAOUT(0,0)
0029      READ (5,905) IANS
0030      IF ( IANS .EQ. 0 ) GOTO 25
C
C  PLOT WEIGHTS ON STRIP CHART
C
0032      DO 160 I=1,NUMW
0033          IF (ABS (WEIGHT (I)) .GT. WMAX ) WMAX= ABS (WEIGHT (I))
0035 160    CONTINUE
0036      WRITE (7,810) WMAX
0037      WSCL=500.0/WMAX
0038      DO 165 I=1, NUMW
0039          WRITE (7,800) I, WEIGHT (I)
0040          TEMP=WEIGHT (I) * WSCL + 0.5
0041          CALL DAOUT (INT (TEMP), 0)
0042 165    CONTINUE
0043      CALL DAOUT (0,0)
C
C  PLOT FREQUENCY RESPONSE OF FILTER (MAGNITUDE AND PHASE)
C
0044      NF=200
0045      CALL FREGMP (WEIGHT, NUMW, NF, REAL, PHA)
0046      RMAX=0.0
0047      PMAX=0.0
0048      RMIN=10000.0
0049      PMIN=10000.0
0050      DO 170 I= 1, NF
0051          IF ( REAL(I) .GT. RMAX ) RMAX=REAL(I)
0052          IF ( REAL(I) .LT. RMIN ) RMIN=REAL(I)
0053          IF ( PHA(I) .GT. PMAX ) PMAX=PHA(I)
0054          IF ( PHA(I) .LT. PMIN ) PMIN=PHA(I)
0055 170    CONTINUE
0056      RSCL=RMIN*(-1.0)
0057      IF ( RMAX .GT. (-1.0)*RMIN ) RSCL=RMAX
0058      PSCL=PMIN*(-1.0)
0059      IF ( PMAX .GT. (-1.0)*PMIN ) PSCL=PMAX
0060      PSCL=500.0/PSCL
0061      RSCL=500.0/RSCL
0062      DO 180 I=1, NF
0063          REAL(I)=(REAL(I)*RSCL)+0.5
0064          N1=INT(REAL(I))
0065          PHA(I)=(PHA(I)*PSCL)+0.5
0066          N2=INT(PHA(I))
0067          WRITE (7,810) REAL(I), PHA(I)
0068          FORMAT (1X,G12.5,1X,G12.5)
0069      CALL DAOUT (N1,N2)
0070 180    CONTINUE
0071      CALL DAOUT (0, 0)

```

C SET UP FOR RERUN IF DESIRED

C

```
078      WRITE (7,900)
079      900  FORMAT (1X, 'ANOTHER GO: 1=NO ?',%)
080      READ (5,905) IANS
081      905  FORMAT (I1)
082      IF (IANS .EQ. 1) GOTO 1000
084      35  WRITE (7,910)
085      910  FORMAT (1X, 'GET NEW INPUTS ?',%)
086      READ (5,905) IANS
087      WRITE (7, 915)
088      915  FORMAT (1X, 'RESET WEIGHT VECTOR ?',%)
089      READ (5,905) IANS1
090      IF (IANS1 .EQ. 1) GOTO 30
092      DO 190 I= 1, NUMW
093          WEIGHT(I)=0
094      190  CONTINUE
095      30  WRITE (7,920)
096      920  FORMAT (1X, 'NEW ADAPTIVE CO-EFFICIENT ?',%)
097      READ (5, 905) IANS1
098      IF (IANS1 .EQ. 1) GOTO 40
099      WRITE (7, 925)
100      925  FORMAT (1X, 'INPUT NEW ADAPTIVE WEIGHT COEFFICIENT :',%)
101      READ (5,930) ADAPT
102      930  FORMAT (G14.7)
103      40  IF (IANS .EQ. 1) GOTO 20
104      GOTO 10
106
```

C

C ACTUAL BEGINNING OF PROGRAM (FIRST RUN)

C

```
0107      50  WRITE (7,935)
0108      935  FORMAT (1X, 'ENTER CHANNEL NUMBER OF NOISE SOURCE :',%)
0109      READ (5,940) NC1
0110      NC1=256*NC1+16
0111      WRITE (7,936)
0112      936  FORMAT (1X, 'ENTER CHANNEL NUMBER OF NOISE+SIGNAL :',%)
0113      READ (7,940) NC2
0114      NC2=256*NC2+16
0115      WRITE (7,937)
0116      937  FORMAT (1X, 'ENTER NUMBER OF SAMPLES :',%)
0117      READ (5,940) NUMS
0118      940  FORMAT (I5)
0119      WRITE (7,950)
0120      950  FORMAT (1X, 'ENTER NUMBER OF DELAYS FOR FILTER :',%)
0121      READ (5,940) NUMD
0122      NUMD=NUMD-1
0123      WRITE (7, 955)
0124      955  FORMAT (1X, 'ENTER NUMBER OF WEIGHTS UPTO 100 :',%)
0125      READ (5,940) NUMW
0126      IANS=0
0127      GOTO 30
0128      1000 CALL DACUT(0,0)
0129      STOP
0130      END
```

DRTRAN IV Storage Map for Program Unit ADAPT

Local Variables, .PSECT \$DATA, Size = 053154 (11062. words)

ame	Type	Offset	Name	Type	Offset	Name	Type	Offset
DAPT	R*4	053052	H	R*4	053048	I	I*2	053016
ANS	I*2	053056	IANS1	I*2	053132	J	I*2	053036
	I*2	053040	NC1	I*2	053022	NC2	I*2	053024
F	I*2	053074	NUMD	I*2	053030	NUMS	I*2	053020
UMW	I*2	053026	N1	I*2	053126	N2	I*2	053130
UT	R*4	053042	RMAX	R*4	053102	RMIN	R*4	053112
SCL	R*4	053122	RMAX	R*4	053076	RMIN	R*4	053106
SCL	R*4	053116	SUM	R*4	053032	TEMP	R*4	053070
MAX	R*4	053060	WSCL	R*4	053064			

Local and COMMON Arrays:

ame	Type	Section	Offset	Size	Dimensions
N1	I*2	\$DATA	000620	023420 (5000.)	(5000)
N2	I*2	\$DATA	024240	023420 (5000.)	(5000)
HA	R*4	\$DATA	051324	001444 (402.)	(201)
EAL	R*4	\$DATA	047660	001444 (402.)	(201)
EIGHT	R*4	\$DATA	000000	000620 (200.)	(100)

Subroutines, Functions, Statement and Processor-Defined Functions:

ame	Type	Name	Type	Name	Type	Name	Type	Name	Type
BS	R*4	DADUT	R*4	FREQMP	R*4	INT	I*2	XTSMP2	R*4

VARIABLE LIST

ADAPT- Adaptive coefficient
H - Product of the filter output and adaptive coefficient
I - Pointer for Do loops
IANS - Variable for responses from the keyboard
IANS1- Variable for responses from the keyboard
J - Pointer for Do loops
K - The present input value with delay accounted for
NC1 - Address pointer for the channel of the noise source
NC2 - Address pointer for the channel of the signal + noise source
NF - Number of frequencies between 0 and $f_s/2$ the frequency spectrum
of the weight vector is calculated
NUMD - Number of delays in the filter
NUMS - Number of samples taken
NUMW - Number of weights in the filter
N1 - Integer value of array REAL scaled between 0 and 500
N2 - Integer value of array PHA scaled between 0 and 500
OUT - Filter output
PMAX - Maximum value of the array PHA
PMIN - Minimum value of the array PHA
PSCL - Factor required to scale array PHA to range from 0 to 500
RMAX - Maximum value of the array REAL
RMIN - Minimum value of the array REAL
RSCL - Factor required to scale the array REAL to range from 0 to 500
SUM - Summation variable for the fixed filter
TEMP - Scaled value of the array WEIGHT
WMAX - Maximum value of the array WEIGHT
WSCL - Factor required to scale the array WEIGHT to range from 0 to 500

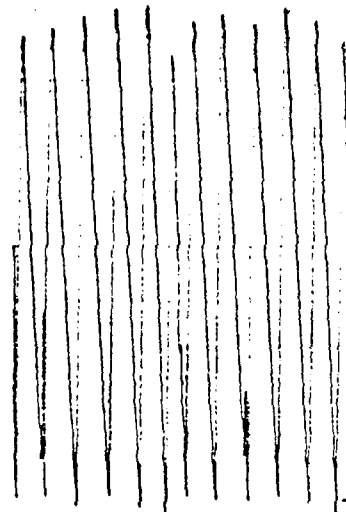
ARRAY LIST

IN1 - Input array containing noise samples (5000)
IN2 - Input array containing signal + noise samples (5000)
PHA - Phase values of the frequency spectrum of the weight vector (201)
REAL - Magnitude values of the frequency spectrum of the weight array (201)
WEIGHT- Weight vector (100)

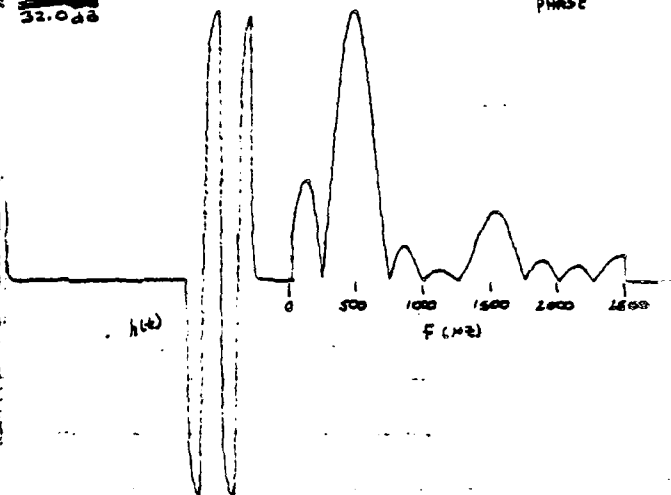
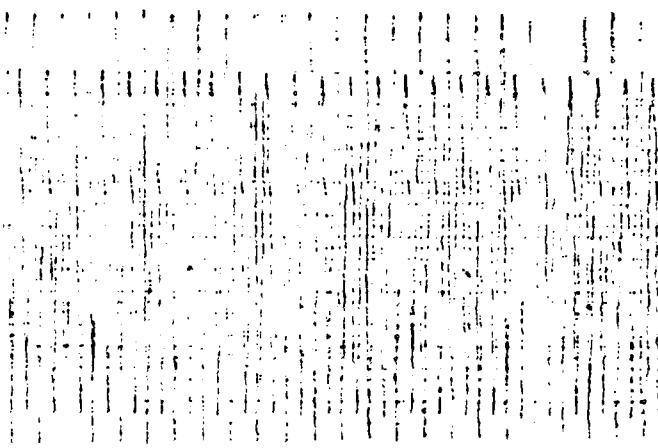
APPENDIX I SAMPLE RUNS



$f_s = 5000 \text{ Hz}$
 $f_{ref} = 500 \text{ Hz}$ SQUARE WAVE
 $f_{ref} = 750 \text{ Hz}$ SQUARE WAVE
 $f_{ref} = 600 \text{ Hz}$ TRIANGULAR WAVE
 $\Delta t = 1.25 \cdot 10^{-8}$ NUMBER OF WEIGHTS = 20
 0 DELAY 750 ADAPTATIONS
 $SNR_{IN} = -18.84 \text{ dB}$ $SNR_{OUT} = 13.00 \text{ dB}$
 $SNR_{IMPROV} = 32.0 \text{ dB}$



PHASE



$f = 700 \text{ Hz}$

0.0

0.0

0.0

0.0

$$W_L(k+1) = W_L(k) + uYS(k-1)$$

where W_L are the N weights, u is the adaptive gain coefficient, Y is the last output of the system, and S are the reference inputs (N_0). The sequence of four instructions that performs this operation are:

```

ZALH *
MPY 68
APAC
SACH *-
ZALH *
MPY 67
APAC
SACH *-
ZALH *
MPY 66
APAC
SACH *-
etc.

```

The ZALH * instruction zeroes the accumulator and loads the contents of the data memory location pointed to by the current auxiliary register (set to point to the weights) into the upper 16 bits of the accumulator. The MPY multiplies the contents of the data memory location (location 68, 67 ... contains the past reference inputs) by the T register (which contains the adaptive gain coefficient and output product uxY). The APAC instruction adds the contents of the P register (result of last multiply) to the accumulator and stores result in the accumulator. The SACH *- instruction stores the upper 16 bits of the accumulator into the data memory location pointed to by the current auxiliary register (still points to the location of the weight) and the auxiliary register is decremented.

The tapped delay line portion of the program calculates:

$$F(k) = \sum_{i=1}^N W_i(k) S(k-i)$$

where F is the output of the adaptive filter, N is the number of weights, S are the past N reference inputs, and W are the N weights. This can be implemented with a sequence of two instructions.

```
LTD 67
MPY *-
LTD 66
MPY *-
LTD 65
MPY *-
etc.
```

The LTD instruction places the contents of the data memory location (locations 67, 66, ... contain the past reference inputs) in the T register to set up for the next multiply, add the result of the last multiply which is in the product register to the accumulator, and shifts the data memory to the next data memory location(contents of location 67 would be placed in 68). The MPY *- will multiply the contents of the T register by the contents of the data memory that the current auxiliary register points to (it is set up to point to the data memory containing the weights) and then decrement the auxiliary register to point to the next weight.

The adaptive filter portion of the program uses a series of four instructions. The new weights are calculated by:

Figure 2 shows a flow diagram of the a TMS-320 program that performs the adaptive noise cancellation. The program begins with an initialization portion to set up certain constants. In the main loop, there are several steps performed. First a reference input (N_0) is read. Next the tapped delay line filter is performed. The output of the system is next computed and transferred to the output port. Then the signal plus noise input ($S + N_1$) is read. Finally, the filter weights are adaptively adjusted. This loop is performed once every sampling period. Delay loops are also part of the loop to control the sampling rate. To show the power of the TMS-320 instruction set, the adaptive filter portions of the program are explained.

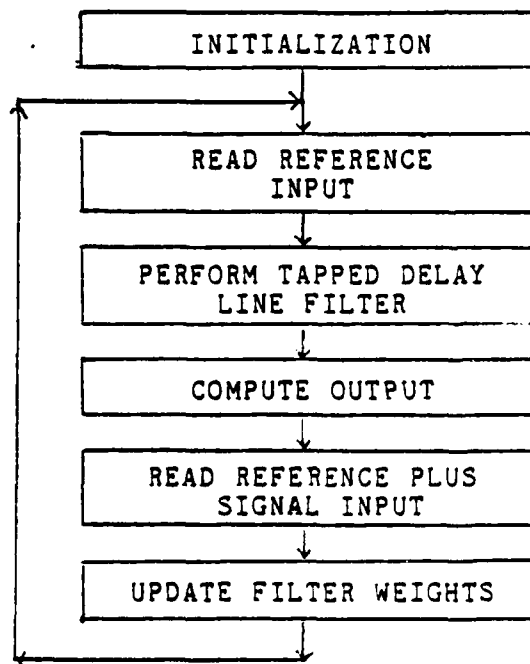


Figure 2. Flow diagram of adaptive noise cancellation program.

Appendix C.

Implementation of an Adaptive Noise Cancellation Algorithm on the TMS-320

An adaptive noise cancellation system has been developed on the Texas Instruments TMS-320. A block diagram of the system is shown in Figure C1. The TMS-320 can implement a 68 weight adaptive filter operating at a sampling frequency of 10.7 kHz.

The TMS-320 has an instruction set tailored to do digital signal processing. With its 200 nsec instruction cycle, it becomes a very powerful processor for applications such as this. The program to do the adaptive noise cancellation only requires six instruction per weight or 1.2 usec plus the overhead needed to do input, output, and minor data manipulations. The maximum number of filter weights that the TMS-320 can support without major hardware modification is 68.

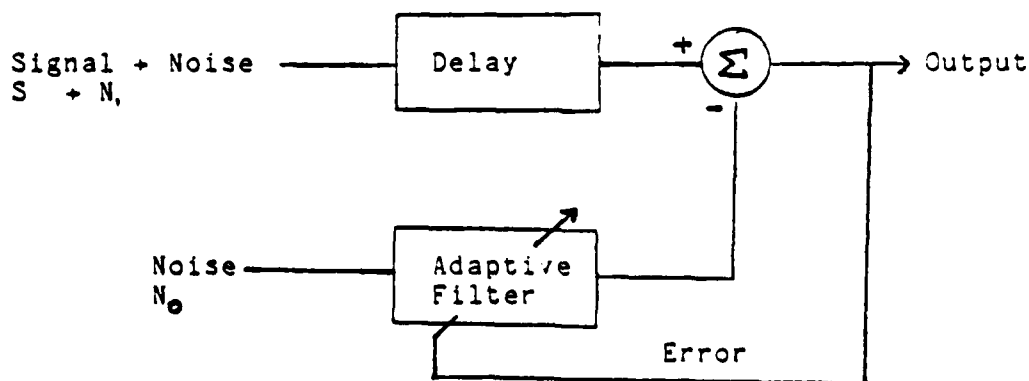


Figure C1. Adaptive Noise Cancellation System.

Enclosure 3 to Appendix B Assembly language sampling subroutine called from
rain program. (Also used with programs in Appendixes E and F.)

```

.TITLE XTSMP2
.GLOBAL XTSMP2
;
;
;      B. B. PETERSON 17 NOV 83
;      SUBROUTINE CALLED BY "CALL XTSMP2(N,N1,N2,IX1,IX2)". N= # OF
;      SAMPLES IN EACH OF CHANNELS NC1 AND NC2, (N1=256*NC1+16 ETC.)
;      AND IX1 AND IX2 ARE THE RETURNED DATA VECTORS.
;      MAXIMUM TRIGGER RATE IS 21,000 SAMPLES/SEC (10,500 PER CHANNEL)
;
      IOC=170400      ;ADDRESS OF A/D CONTROL REGISTER
      IOI=170402      ;ADDRESS OF A/D OUTPUT REGISTER
XTSMP2: TST (R5)+
      MOV #(R5)+,R0    ;R0 = # OF SAMPLES
      MOV #(R5)+,R1    ;R1 = N1
      MOV #(R5)+,R2    ;R2 = N2
      MOV (R5)+,R3     ;R3 = STARTING ADDRESS OF DATA VECTOR IX1
      MOV (R5)+,R4     ;R4 = STARTING ADDRESS OF IX2
      MOV R1,#IOC      ;ENABLE EXTERNAL TRIGGER, SET ADD TO CH NC1
LOOP:   BIT #200,#IOC   ;TEST DONE BIT
      BEQ LOOP
      MOV #IOI,R5      ;RESET DONE BIT, THROW AWAY DATA
TEST1:  BIT #200,#IOC   ;TEST DONE BIT
      BEQ TEST1        ;WAIT AND TEST AGAIN IF NOT SET
      MOV R2,#IOC      ;SET ADDRESS TO CH NC2
      MOV #IOI,(R3)+   ;READ A/D AND PUT IN MEMORY
LOOP2:  BIT #200,#IOC   ;TEST DONE BIT
      BEQ LOOP2
      MOV R1,#IOC      ;SET ADDRESS TO CH NC1
      MOV #IOI,(R4)+   ;READ CH NC2
      DEC R0
      BNE TEST1        ;CHECK TO SEE IF N SAMPLES HAVE BEEN TAKEN
      RTS PC           ;IF SO, RETURN
      .END

```

Enclosure 2 to Appendix B

BY G. B. PETERSON & NOV 82

SUBROUTINE FOR SOLVING SIMULTANEOUS LINEAR ALGEBRAIC
EQUATIONS USING GAUSS ELIMINATION WITH ROW PIVOTING.

FROM "CIRCUIT THEORY: A COMPUTATIONAL APPROACH" BY G. W.
DIRECTOR, PAGE 211.

FORM OF CALL IS:

CALL GAUSS

THE N*N MATRIX A, THE N VECTOR B, THE N VECTOR X AND THE DIMENSION
N MUST BE COMMON VARIABLES.

SUBROUTINE GAUSS

COMMON A(44,44),B(44),N

DO 5 I=1,N

I1=I+1

IF (ABS(A(I,I)).LE.1.E-10) GO TO 1

GO TO 15

CONTINUE

IF (I.EQ.N) GO TO 10

DO 14 J=I1,N

IF (ABS(A(J,I)).LE.1.E-10) GO TO 14

IPIV=J

GO TO 16

CONTINUE

GO TO 10

DO 2 K=I1,N

PIV=A(IPIV,K)

A(IPIV,K)=A(I,K)

A(I,K)=PIV

PIV=B(IPIV)

B(IPIV)=B(I)

B(I)=PIV

IF (I.EQ.N) GO TO 2

DO 8 J1=I1,N

A(I,J1)=A(I,J1)/A(I,I)

B(I)=B(I)/A(I,I)

DO 5 J=I1,N

DO 4 K=I1,N

A(J,K)=A(J,K)-A(J,I)*A(I,K)

B(J)=B(J)-B(I)*A(I,K)

B(N)=B(N)/A(N,N)

DO 6 K=2,N

I=N-K+1

DO 7 J=I,N

SUM=SUM+A(I,J)*B(J)

B(I)=B(I)-SUM

GO TO 11

WRITE(7,9)

FORMAT(' EQUATIONS ARE LINEARLY DEPENDENT')

STOP

RETURN

END

Enclosure 1 to Appendix B

```
1925      DO 1925 ILOOP= 1 ,NLOOP
C          CONTINUE
C          PLOT OUT DATA
C          CALL DAOUT (IX1(I+N/2),INT(OUT))
1950      FORMAT (2I7,F10.2)
2000      CONTINUE
      WRITE (7,2100) AVG/(NS-NT+1),AIN/AVG
2100      FORMAT (' AVERAGE OUTPUT SQUARED',F14.2,' SNR IMPROVEMENT',F8.2)
      WRITE (7,2200)
2200      FORMAT (' NEW DATA, SAME WEIGHTS? 1=YES, 0= NO')
      READ (5,2300) IFLAG
2300      FORMAT (I6)
      IF (IFLAG.EQ.0) GO TO 1050
      CALL XTSMPZ(N,NC1,NC2,IX1,IX2)
      GO TO 1600
3000      STOP
      END
```

Enclosure 1. to Appendix B

```

1000  CONTINUE
1050  WRITE (7,1100)
1100  FORMAT (' NUMBER OF WEIGHTS,NUMBER OF DELAY LOOPS')
      READ (5,1200) N,NLOOP
1200  FORMAT(2I6)
      IF (N.GT.70) GO TO 3000
C
C  CALCULATE A & B FOR GAUSS SUBROUTINE
C
      DO 1300 J=1,N
          DO 1225 K=1,N
              IF (IFLAG.EQ.0) GO TO 1240
              ID=IABS(J-K)+1
              A(J,K)=C(ID)
1225  WRITE (7,1230) J,K,A(J,K)
1235  FORMAT (' A(',I2,',',I2,',') = ',E12.5)
1240  B(J)=0.0
          DO 1250 I=1,NS+1-NT
              B(J)=B(J)+FLOAT(IX2(I+J-1))*FLOAT(IX1(I+N/2))
1250  IF (IFLAG.EQ.0) GO TO 1300
              WRITE (7,1280) J,B(J)
1280  FORMAT (' B(',I2,',') = ',E12.5)
1300  CONTINUE
C
C  USE GAUSS ELIMINATION SUBROUTINE TO CALCULATE OPTIMUM WEIGHTS
C
      CALL GAUSS
C
      PRINT OUT WEIGHTS
C
      DO 1400 J=1,N
          WRITE (7,1500) J,B(J)
1400  CONTINUE
1500  FORMAT (I6,F13.7)
1600  AVG=0.0
C
C  CALCULATE FILTER OUTPUT USING OPTIMUM WEIGHTS
C
      DO 2000 I=1,NS+1-NT
          OUT=FLOAT(IX1(I+N/2))
          DO 1900 J=1,N
              OUT=OUT-B(J)*IX2(I+J-1)
1900  CONTINUE
C
C  CALCULATE OUTPUT POWER
C
      AVG=AVG+OUT**2
C
C  PRINT OUT INPUT AND OUTPUT DATA
C
      WRITE (7,1950) I,IX1(I+N/2),OUT
C
C  DELAY SO THAT STRIP CHART RECORDER CAN KEEP UP
C

```

Enclosure 1 to Appendix B

PROGRAM OPTFIL

B. B. PETERSON 10 NOV 83

PROGRAM READS IN TWO CHANNELS OF ANALOG DATA USING SUBROUTINE
XTSMP2. THE OPTIMUM WEIGHTS OF FIR FILTERS OF VARIOUS LENGTHS
ARE CALCULATED USING A ONE SHOT LINEAR LEAST SQUARES TECHNIQUE.
THE AVERAGE POWER IN THE OUTPUT IS THEN CALCULATED FOR EACH LENGTH.

ARRAYS:

IX1 & IX2	INPUT DATA FROM TWO MICROPHONES
A & B	MATRIX AND VECTOR IN SOLUTION OF
	SIMULTANEOUS EQUATIONS. (B IS BOTH THE
	PASSED VECTOR AND THE RETURNED ANSWER
	IN THE SOLUTION OF $AX=B$)
C	AUTOCORRELATION VECTOR OF THE REFERENCE
	INPUT (IX2) USED TO CALCULATE A.

INTEGER IX1(3000),IX2(3000)
COMMON A(50,50),B(50),N
REAL C(50)

INPUT PARAMETERS

```

WRITE (7,100)
100  FORMAT (' ENTER # SAMPLES,S+N CH,NOISE CH,LARGEST FILTER LENGTH')
    READ (5,400) NS,NC1,NC2,NT
400  FORMAT (4I7)
    WRITE (7,500)
500  FORMAT (' TYPE DATA? (1=YES,0=NO)')
    READ (5,600) IFLAG
600  FORMAT (I4)
    AIN=0.0
    DO 700 J=1,NT
650      C(J)=0.0
700  CONTINUE

```

DELAY BEFORE SAMPLING

```

DO 850 I=1,10000
850  CONTINUE
    N1=256*NC1+16
    N2=256*NC2+16
    CALL XTSMP2(NS,N1,N2,IX1,IX2)
    IF (IFLAG.EQ.0) GO TO 940
900  DO 920 I=1,NS
920      WRITE (7,930) I,IX1(I),IX2(I)
930  FORMAT (3I7)
940  DO 1000 I=1,NS+1-NT
950      DO 990 J=1,NT
          C(J)=C(J)+FLOAT(IX2(I+J-1))*FLOAT(IX2(I))
990  CONTINUE
    CALCULATE INPUT POWER
    AIN=AIN+FLOAT(IX1(I))*FLOAT(IX1(I))

```

However, in the program (enclosure 1) because of the long time required for the large number of calculations, they were set equal which allowed the calculation of only one autocorrelation vector (C). The longer the data set the more valid this approximation is.

After the A matrix and the B vector are calculated, the optimum weight vector is calculated using a Gauss elimination subroutine (enclosure 2). The filter output was then calculated using this weight vector and the output energy compared to the primary input energy.

The results were not impressive. Data from both the Luder yawl engine and the gasoline research were analyzed with minimal reduction in noise power. In view of these results and others obtained since, the explanation of the poor performance of the adaptive early filters, was there limited length and not necessarily slow convergence.

Repeated versions of equation B1 can be written in matrix form:

$$Y = P - RH \quad (B2)$$

where $P(k) = x_1(k-N/2)$ and $R_{ki} = x_2(k-i)$. The total output energy is given by:

$$E = Y^T Y = P^T P - 2 H^T R^T P + H^T R^T R H \quad (B3)$$

This energy is minimized by setting it's gradient with respect to H equal to 0.

$$\nabla_H E = -2 R^T P + 2 R^T R H = 0 \quad (B4)$$

$$\text{or } AH = B$$

where $A = R^T R$ and $B = R^T P$.

The elements of A and B are given by:

$$A_{ij} = A_{ji} = \sum_{\text{Data Set}} x_2(k-i) x_2(k-j) \quad (B5)$$

$$B_j = \sum_{\text{Data Set}} x_2(k+j) x_1(k+N/2)$$

Essentially A is an $N \times N$ autocorrelation matrix of the reference input and B is a N vector of the cross-correlation of the reference input and the delayed primary input. Strictly speaking $A_{11} \neq A_{22}$ and $A_{12} \neq A_{23}$ etc. because A_{11} contains one early sample not in A_{22} and A_{22} contains a late sample not in A_{11} .

Appendix B

FORTTRAN Program to Calculate Optimum Filter Weights

In the LMS adaptive noise cancellation scheme the weights of the adaptive filter are adjusted to minimize the output power. In previous FORTRAN implementations of the LMS algorithm, the filter was found to be ineffective in canceling engine noise. It was felt this may be due to the weights not converging within the space of the data set. Using off-line parameter estimation techniques it is possible to calculate the optimum fixed weight filter for a given data set, and then analyze this filter using the data set. The filter is optimum in the sense that the remaining output power for the data set is the minimum of all possible weight vectors. The filter is illustrated in Figure B1.

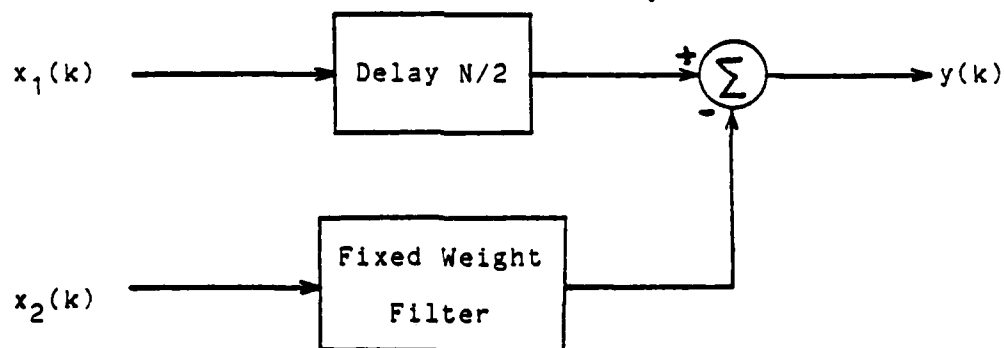
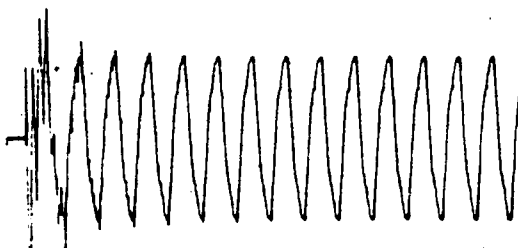


Figure B1

For a fixed weight vector H , the filter output, $y(k)$ is given by:

$$y(k) = x_1(k-N/2) - \sum_{i=0}^{N-1} h_i x_2(k-i) \quad (B1)$$



$f_s = 5000 \text{ Hz}$

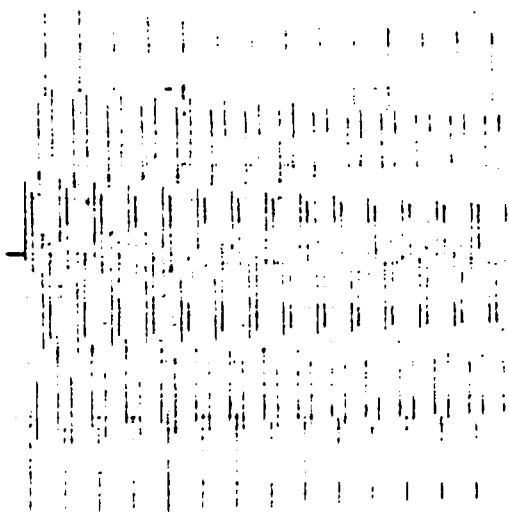
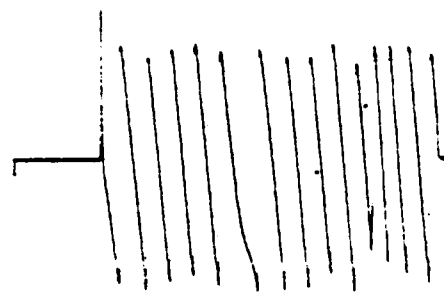
$f_{\text{sig}} = 500 \text{ Hz}$ SQUARE WAVE

$f_{\text{mod}} = 100 \text{ Hz}$ TRIANGLE WAVE

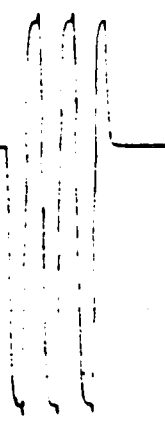
NOISE - 500 Hz TRIANGLE WAVE

$\mu = 1.25 \cdot 10^{-8}$ NUMBER OF WEIGHTS = 30

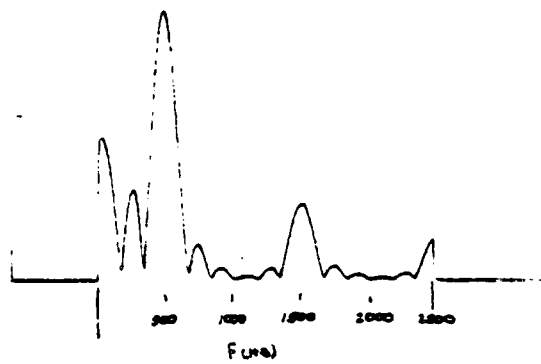
0 DELAY 750 ADAPTATIONS



A22



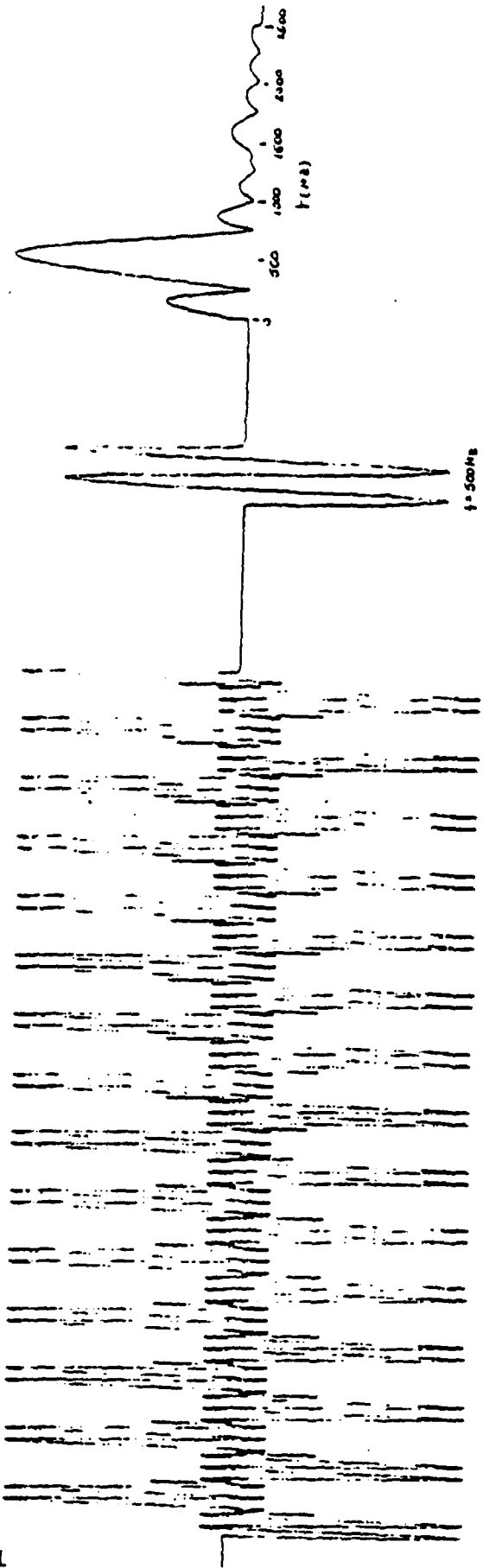
$f = 100 \text{ Hz}$



Freq

$f_s = 5000 \text{ Hz}$
 $f_{\text{ref}} = 500 \text{ Hz}$ SQUARE WAVE
 $f_{\text{m}} = 100 \text{ Hz}$ SQUARE WAVE
 $\text{noise} = 500 \text{ Hz}$ SINE WAVE
 $M = 10 \cdot 10^{-9}$ NUMBER OF WEIGHTS = 20
 $\text{DELAY} = 760$ ADAPTATIONS
 $\text{SNR}_{\text{in}} = -12.3 \text{ dB}$ $\text{SNR}_{\text{out}} = 7.3 \text{ dB}$

IMPROVEMENT - 21.7 dB



The monitor on the TMS-320 development system allows assembly language programs to be down-loaded from another computer. It then assembles the program into its machine code for execution. To analyze the effects of various numbers of weights, sampling frequencies, and adaptive gain coefficients would require many modifications to the assembly language program. To make these modifications easy, a BASIC 7 program was developed on the Dartmouth Time Sharing system to generate TMS-320 assembly code. The program is interactive in that it asks for the number of weights, sampling frequency, and adaptive gain coefficient. It produces straight line code so the time spent on the tapped delay line filtering and weight adaptation is minimized. There are delay loops added to the code to provide a range of sampling frequencies. Enclosure C1 is a listing of the Basic program. Enclosure C2 is a listing of the assembly language code produced by this program for a 10 weight filter.

This system has been implemented to cancel engine background noise in voice communications. Table C1 summarizes the amount of noise reduction for various numbers of weights, sampling frequencies, and adaptive gain coefficients. The amount of noise reduction is calculated by comparing the RMS voltage of the signal plus noise input to the RMS voltage of the output. Anti-aliasing filters set at half the sampling frequency were used on both input signals.

From the data shown in Table C1, it is clear that the amount of noise reduction is improved by increasing the number of filter weights. At the higher sampling rates, there is very little noise reduction. At the lower sampling frequencies the noise reduction is even more apparent (8 dB at 2 kHz.). This indicates that to achieve the same amount of noise reduction at a sampling frequency of 10 kHz would require five times more weights or on the order of 300 weights. Also there is an apparent increase of signal reduction when the adaptive gain coefficient is increased. This is not only an increase in noise reduction; but, because the filter is adapting so quickly to the signal, it is also canceling some of the desired signal and not just the noise.

The noise cancellation system only performed marginally in the engine noise application at useful sampling frequencies. To improve the performance would require a processor capable of implementing more weights. The next generation of TMS-320 may have this capability. This work has shown that adaptive filtering with a maximum of 68 weights is realizable for speech processing and may have other applications.

Sampling Frequency KHz	Number of Weights	Adaptive Gain	Signal Reduction dB
10	68	14	3.5
10	68	12	2.6
10	68	10	2.4
10	68	8	1.4
10	68	6	*
10	50	14	3.2
10	50	12	2.3
10	50	10	2.3
10	50	8	1.5
10	50	6	*
10	30	14	2.5
10	30	12	1.9
10	30	10	1.9
10	30	8	1.5
10	30	6	*
5	68	14	2.4
5	68	12	2.4
5	68	10	2.1
5	68	8	- .8
5	68	6	*
5	50	14	2.1
5	50	12	2.2
5	50	10	2.2
5	50	8	0.0
5	50	6	*
5	30	14	1.9
5	30	12	1.9
5	30	10	1.9
5	30	8	0.9
5	30	6	*
2	68	14	8.4
2	68	12	7.9
2	68	10	6.0
2	68	8	0.6
2	68	6	*
2	50	14	6.4
2	50	12	5.6
2	50	10	4.9
2	50	8	0.4
2	50	6	- .4
2	30	14	2.4
2	30	12	2.4
2	30	10	2.1
2	30	8	- .8
2	30	6	*

(* - Did not track)

Table C1. Test results of Noise Cancellation System.

Enclosure 1 to Appendix C

```
100 !
110 !   PROGRAM NAME: ADAPT-F (ADAPTIVE FILTER)
120 !           VERSION 4
130 !
140 !   PROGRAMMER: K. U. DYKSTRA
150 !
160 !   DATE: 20 SEPT 84
170 !
180 !   THIS PROGRAM PRODUCES THE ASSEMBLY LANGUAGE ADAPTIVE
190 !   FILTER PROGRAM FOR THE TMS-320 EVALUATION MODULE. IT
200 !   CAN BE USED IN CONJUNCTION WITH THIS MODULE TO DOWN
210 !   LOAD THE ASSEMBLY LANGUAGE PROGRAM. A TAPPED DELAY
220 !   LINE ALGORITHM IS USED FOR THE ACTUAL FILTERING. THE
230 !   WEIGHTS ARE ADJUSTED USING A LMS ALGORITHM. INPUTS
240 !   ARE NUMBER OF WEIGHTS, SAMPLING FREQ, ADAPTIVE
    COEFFICIENT.
250 !   THE CODE THAT IS PRODUCED IS STRAIGHT LINE CODE WITH
260 !   NO LOOPING FOR MAXIMUM SPEED.
270 !
280 !   DATA MEMORY ASSIGNMENT
290 !
300 !   0-68  REFERENCE INPUTS
310 !   69    CHANNEL 1 CONTROL
320 !   70    CHANNEL 2 CONTROL
330 !   71    MASK FOR INPUTS
340 !   72    MASK FOR OUTPUTS
350 !   73    FILTER OUTPUT
360 !   74    SIGNAL + REFERENCE INPUT
370 !   75    OUTPUT
380 !   76-143 FILTER WEIGHTS
390 !
400 !   INPUT NUMBER OF WEIGHTS.
410 !
412 DO
414 LET ERROR = 0
420 PRINT "ENTER NUMBER OF WEIGHTS (1-68). "
430 INPUT NWTS
432 IF NWTS < 1 OR NWTS > 68 THEN LET ERROR = 1
434 LOOP UNTIL ERROR = 0
440 !
450 !   INPUT SAMPLING FREQUENCY
460 !
470 LET LOWF=1/((.0002*(37+(6*NWTS)+(20*256)))
472 LET HIGHF=1/((.0002*(37+(6*NWTS)+20))
474 IF HIGHF > 15.6 THEN LET HIGHF = 15.6
476 DO
478 LET ERROR = 0
480 PRINT "ENTER SAMPLING FREQUENCY (";LOWF; " TO ";HIGHF;"
    KHZ)."
490 INPUT FREQ
495 IF FREQ < LOWF OR FREQ > HIGHF THEN LET ERROR = 1
496 LOOP UNTIL ERROR = 0
```

```

500 LET DELAY=INT(((1/(FREQ*.0002))-33-(6*NWTS))/20)
510 LET AFREQ=1/((.0002*(37+(6*NWTS)+(20*DELAY)))
520 PRINT "SAMPLING FREQUENCY = ";AFREQ;" DELAY = ";DELAY
530 !
532 DO
534 LET ERROR = 0
540 PRINT "ADAPTIVE COEFFICIENT = 1/(2**(16-N)), ENTER N"
550 PRINT "IF YOU WANT TO DOWN LOAD END WITH <CTRL C> ELSE
    <RETURN>"
560 INPUT COEF
562 IF COEF < 0 OR COEF > 15 THEN LET ERROR = 1
564 LOOP UNTIL ERROR = 0
570 !
580 !     SET UP DATA CONSTANTS
590 !
600 PRINT ">"
610 PRINT "      AORG 0"
620 PRINT "      B BEGIN"
630 PRINT "      NOP"
640 PRINT "      NOP"
650 PRINT "      DATA >7FF0"
660 PRINT "      DATA >8000"
670 PRINT "BEGIN LACK 4"
680 PRINT "      TBLR 71"
690 PRINT "      LACK 5"
700 PRINT "      TBLR 72"
710 PRINT "      LACK 3"
720 PRINT "      SACL 69"
730 PRINT "      LACK >83"
740 PRINT "      SACL 70"
750 PRINT "      SOVM"
760 PRINT "      LARP 1"
770 !
780 !     INPUT REFERENCE INPUT
790 !
800 PRINT "LOOP  OUT 69,0"
810 PRINT "      NOP"
820 PRINT "      NOP"
830 PRINT "      IN 0,2"
840 PRINT "      ZALS 71"
850 PRINT "      XOR 0"
860 PRINT "      SACL 0"
870 !
880 !     DELAY FOR SAMPLING FREQUENCY
890 !
900 PRINT "      LARP 0"
910 PRINT "      LARK 0,";DELAY
920 PRINT "D1  NOP"
921 FOR I = 1 TO 7
922 PRINT "      NOP"
923 NEXT I
930 PRINT "      BANZ D1"
940 !
950 !     PERFORM TAPPED DELAY LINE FILTER

```

```

960 !
970 PRINT "      ZAC"
975 PRINT "      MPYK 0"
980 PRINT "      LARP 1"
990 PRINT "      LARK 1,143"
1000 FOR I = NWTS - 1 TO 0 STEP -1
1010     PRINT "      LTD ";I
1020     PRINT "      MPY *-"
1030 NEXT I
1040 PRINT "      APAC"
1050 PRINT "      SACH 73"
1060 !
1070 !     OUTPUT = SIGNAL&REF - FILTERED OUTPUT
1080 !
1090 PRINT "      LAC 74"
1100 PRINT "      SUB 73"
1110 PRINT "      SACL 75"
1120 !
1130 !     COEF*OUTPUT > T
1140 !
1150 PRINT "      LAC 75,";COEF
1160 PRINT "      SACH 73"
1170 PRINT "      LT 73"
1180 !
1190 !     OUTPUT
1200 !
1210 PRINT "      ZALS 72"
1220 PRINT "      XOR 75"
1230 PRINT "      SACL 75"
1240 PRINT "      OUT 75,2"
1250 !
1260 !     INPUT REFERENCE&SIGNAL
1270 !
1280 PRINT "      OUT 70,0"
1290 PRINT "      NOP"
1300 PRINT "      NOP"
1310 PRINT "      IN 74,2"
1320 PRINT "      ZALS 71"
1330 PRINT "      XOR 74"
1340 PRINT "      SACL 74"
1350 !
1360 !     DELAY FOR SAMPLING FREQUENCY
1370 !
1380 PRINT "      LARP 0"
1390 PRINT "      LARK 0,";DELAY
1400 PRINT "D2      NOP"
1401 FOR I = 1 TO 7
1402 PRINT "      NOP"
1403 NEXT I
1410 PRINT "      BANZ D2"
1420 PRINT "      LARP 1"
1430 PRINT "      LARK 1,143"
1440 !
1450 !     PERFORM ADAPTIVE PART

```

```
1460 !
1470 FOR I = NWTS-1 TO 0 STEP -1
1480     PRINT "      ZALH *"
1490     PRINT "      MPY ";I+1
1500     PRINT "      APAC"
1510     PRINT "      SACH *-"
1520 NEXT I
1530 PRINT "      B LOOP"
1540 PRINT "      END"
1550 PRINT "<"
1560 END
```

Enclosure 2 to Appendix C

ENTER NUMBER OF WEIGHTS (1-68).

? 10

ENTER SAMPLING FREQUENCY (.958405 TO 15.6 KHZ).

? 10.0

SAMPLING FREQUENCY = 10.0604 DELAY = 20

ADAPTIVE COEFFICIENT = $1/(2^{16-N})$, ENTER N

IF YOU WANT TO DOWN LOAD END WITH <CTRL C> ELSE <RETURN>

? 12

>

```
      AORG 0
      B BEGIN
      NOP
      NOP
      DATA >7FF0
      DATA >8000
BEGIN, LACK 4
      TBLR 71
      LACK 5
      TBLR 72
      LACK 3
      SACL 69
      LACK >83
      SACL 70
      SOVM
      LARP 1
LOOP, OUT. 69,0
      NOP
      NOP
      IN 0,2
      ZALS 71
      XOR 0
      SACL 0
      LARP 0
      LARK 0, 20
D1,  NOP
      NOP
      NOP
      NOP
      NOP
      NOP
      NOP
      NOP
      NOP
      BANZ D1
      ZAC
      MPYK 0
      LARP 1
      LARK 1,143
      LTD 9
      MPY *-
      LTD 8
      MPY *-
      LTD 7
```


MPY *-
LTD 6
MPY *-
LTD 5
MPY *-
LTD 4
MPY *-
LTD 3
MPY *-
LTD 2
MPY *-
LTD 1
MPY *-
LTD 0
MPY *-
APAC
SACH 73
LAC 74
SUB 73
SACL 75
LAC 75, 12
SACH 73
LT 73
ZALS 72
XOR 75
SACL 75
OUT 75,2
OUT 70,0
NOP
NOP
IN 74,2
ZALS 71
XOR 74
SACL 74
LARP 0
LARK 0, 20
NOP
NOP
NOP
NOP
NOP
NOP
NOP
NOP
BANZ D2
LARP 1
LARK 1, 143
ZALH *
MPY 10
APAC
SACH *-
ZALH *
MPY 9
APAC

D2

SACH *-
ZALH *
MPY 8
APAC
SACH *-
ZALH *
MPY 7
APAC
SACH *-
ZALH *
MPY 6
APAC
SACH *-
ZALH *
MPY 5
APAC
SACH *-
ZALH *
MPY 4
APAC
SACH *-
ZALH *
MPY 3
APAC
SACH *-
ZALH *
MPY 2
APAC
SACH *-
ZALH *
MPY 1
APAC
SACH *-
B LOOP
END

<

Appendix D.

Notch Filter Analysis and Results

In [1] and [2] it was pointed out that when the reference input ($x_2(k)$ in Figure D1) is periodic in the length N , of the adaptive filter, the result is a notch filter from $x_1(k)$ to $y(k)$.

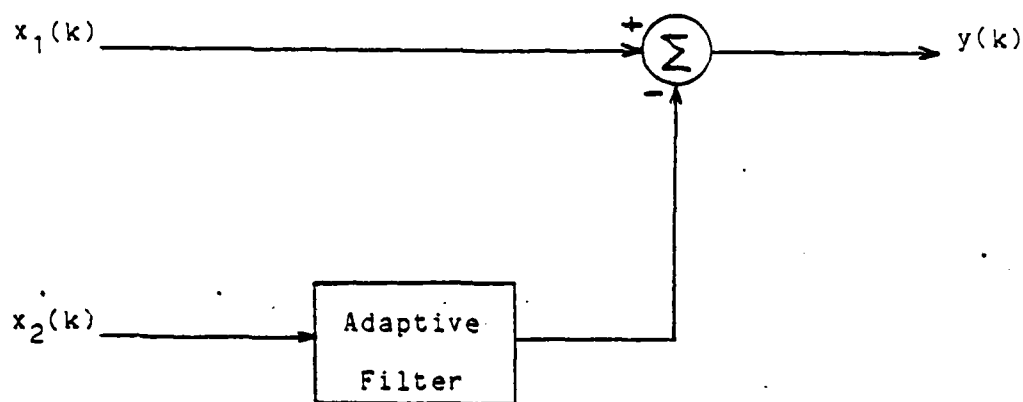


Figure D1

Because the background noise due to an engine is quite periodic, it follows that any reference that contains all the harmonics of the engine noise should suffice. A periodic reference input consisting of one "1" followed by $N-1$ "0"'s i.e.

$$\begin{aligned} x_2(k) &= 1 && \text{for } k = mN \\ &= 0 && \text{for } k = mN \end{aligned} \quad (D1)$$

where N is both the number of samples per period of the engine noise and the number of filter weights and m is an integer not

only contains all these harmonics but also makes possible implementing a filter of large N in real time.

The filter and adaptive algorithm equations now become:

$$\begin{aligned}
 y(mN+i) &= x_1(mN+i) - \sum_{j=0}^{N-1} h_j(mN+i) x_2(mN+i-j) \quad (D2) \\
 &= x_1(mN+i) - h_1(mN+i)
 \end{aligned}$$

and

$$h_1((m+1)N+i) = h_1(mN+i) + a y(k) \quad (D3)$$

Because the algorithm requires only one subtraction, one addition and one multiplication per sample period, it can be implemented very efficiently with the order, N , limited only by memory. Further, if $a = 2^{-n}$ the multiply can be accomplished with right shifts allowing implementation on a general purpose microprocessor.

Equations (D2) and (D3) can be written using z transforms:

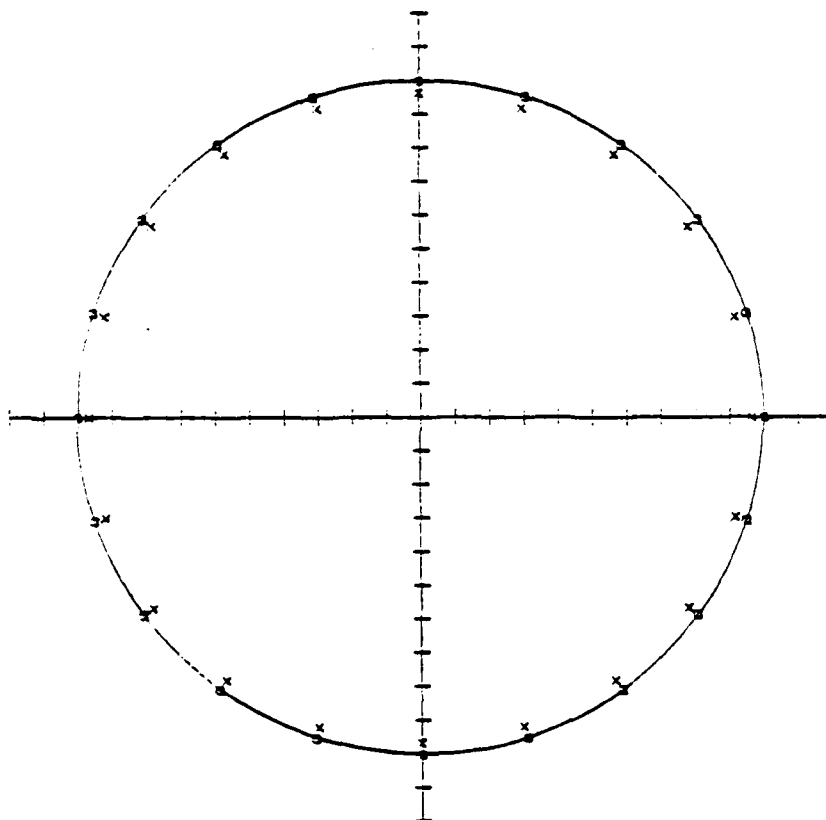
$$Y(z) = X_1(z) - H_1(z) \quad (D4)$$

$$z^N H_1(z) = H_1(z) - a Y(z) \quad (D5)$$

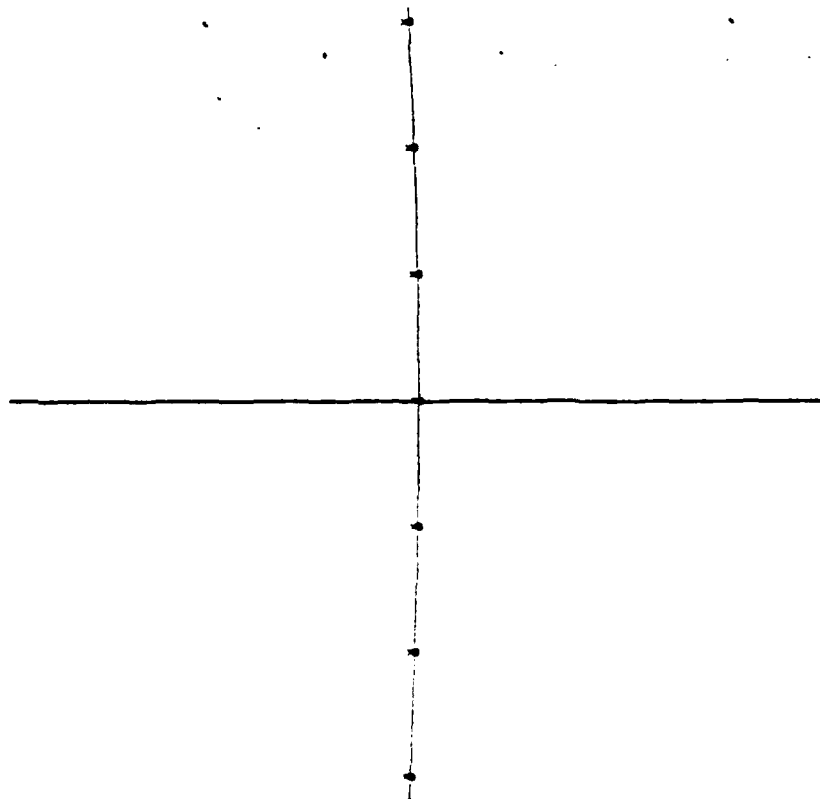
Combining (D4) and (D5) results in the transfer function from $x_1(k)$ to $y(k)$:

$$\frac{Y(z)}{X_1(z)} = \frac{z^N - 1}{z^N - (1-a)} \quad (D6)$$

The N zeros of (D6) are on the unit circle at $e^{j2\pi k/N}$, $k=0$ to $N-1$ and each zero has a corresponding pole at $(1-a)^{1/N} e^{j2\pi k/N}$. A pole-zero plot for $a = 0.5$ and $N = 20$ is shown in Figure D2. The actual filter implemented had $a = 0.25$ and $N = 400$. Each pole



Pole Zero Plot for $N = 20$ and $a = 0.5$
Figure D2



Portion of Pole Zero Plot for $N = 400$ and $a = 0.25$
Figure D3

was therefore at a radius of 0.999281. Figure D3 is a small section of the pole-zero plot for this case illustrating the seven poles and zeros nearest $z=1$. The net effect is a notch filter with a notch at each harmonic of $x_2(k)$. The width (distance between -3 db points) of each notch is approximately:

$$\frac{a f_s}{\pi N} = \frac{a f_o}{\pi}$$

where f_s and f_o are the sampling and fundamental frequencies respectively. For $N = 400$, $a = 0.25$ and $f_s = 7$ kHz, the notches are 1.4 Hz wide and adjacent notches are separated by 17.5 Hz. This frequency response is impossible to verify using our analog spectrum analyzers. Figure D4 shows the output spectrum for $N = 160$, $a = 0.25$ and $f_s = 40$ kHz. The input was a sinusoid swept from 20 Hz to 10 kHz. The notches are 19.9 Hz wide and spaced at 250 Hz intervals. The slight rolloff at high frequencies is due to the hold function of the D/A converter.

If the output of the filter is taken at the output of the adaptive filter before the summer, the resulting filter is a recursive comb filter passing only those components of $x_1(k)$ that are periodic in N samples. Similar non-recursive comb filters that exploited the periodicity of the speech waveform to eliminate white background noise were proposed in [3] and [4]. These filters were adaptive in the sense that the period of the filter was continuously adapted to match the period of the speech waveform. This type of filter is also the digital version of the analog waveform eductors of many years ago.

If $x_2(k)$ is an arbitrary periodic signal with a period of N

Enclosure 1 to Appendix E

PROGRAM ADFDL5

B. B. PETERSON

PROGRAM TO IMPLEMENT ADAPTIVE FILTER IN FREQUENCY DOMAIN
USING WEIGHTED LEAST SQUARES ADAPTIVE ALGORITHM

REF: M. DENTINO, J. MCCOOL, & B. WIDROW, "ADAPTIVE FILTERING
IN THE FREQUENCY DOMAIN", PROC IEEE, VOL.66, NO. 12, DEC 78.
MAXIMUM SIZE OF EACH BLOCK OF DATA IS 512

DIMENSION X1(257),Y1(257),X2(257),Y2(257),X0(257),Y0(257)
DIMENSION WR(512),WI(512),XH(256),YH(256),P2(256),CR(256)
DIMENSION CI(256)
INTEGER I1(5120),I2(5120)

VARIABLES

X1 & Y1 PRIMARY INPUT (BOTH TIME AND FREQUENCY DOMAIN DATA)
X2 & Y2 REFERENCE INPUT
X0 & Y0 OUTPUT
XH & YH WEIGHTS OF ADAPTIVE FILTER (FREQUENCY DOMAIN)
WR & WI TABLES OF COSINES AND SINES
I1 TIME DOMAIN PRIMARY INPUT
I2 TIME DOMAIN REFERENCE INPUT

CALCULATE VECTORS OF SINES AND COSINES

NMAX=512
TPCN=6.28318/FLOAT(NMAX)
DO 3 I=1,NMAX
 PH=FLOAT((I-1))*TPCN
 WR(I)=COS(PH)
 WI(I)=-SIN(PH)

CONTINUE

INPUT PARAMETERS

WRITE (7,10)
FORMAT (' NEW DAT(1=Y),ZER NT,# SAMP,BLK SZ,CH #S,ADP GAIN')
READ (5,20) IF1,IF2,N,MW,NC1,NC2,G
FORMAT (6I6,F10.8)
NW=2*+MW
OMG=1.0-G
IF (IF1.NE.1) GO TO 100
N1 = 256*NC1 + 16
N2 = 256*NC2 + 16

INSERT DELAY BEFORE SAMPLING

DO 30 J=1,20000
CONTINUE
CALL XTSMPC (N,N1,N2,I1,I2)
NW2 = NW/2
IF (IF2.EQ.0) GO TO 50

results are subtracted from $X_1(f)$, giving the FFT of the output, which is then inverse transformed to realize the time domain output. The data is processed in blocks of $N (=2n)$ points.

Since each complex element of $Y(f)$ is a function only of the corresponding coefficient of $H(f)$ and not the entire vector the adaptive algorithm can be made to have much more efficient and predictable convergence properties. Therefore, unlike the time domain problem, when doing off-line processing on the LSI-11/2 with long filter lengths but limited data storage, in the frequency domain filter one can be assured the weights will converge within the data set. Also, because of the efficiency of the algorithm, the data can be processed in reasonable time.

In the FORTRAN implementation (enclosure 1) the complex LMS algorithm proposed in [6] has been changed to a weighted least squares algorithm for more efficient and predictable convergence properties. Each element of $H(f)$ is the best least squares estimate with the input data exponentially weighted. The filter can therefore track time varying parameters with the time constant of the exponential weighting.

Appendix E.

Frequency Domain Adaptive Filtering

A fundamental problem in both real time implementation and in off-line analysis of adaptive noise cancellation is that the filter length is severely limited by memory, processor speed, or both. At the expense of much more complicated software, digital filters, including adaptive filters, can be implemented more efficiently in the frequency domain than the time domain. The basic structure of a frequency domain adaptive noise canceler, proposed in [5] is shown in Figure E1.

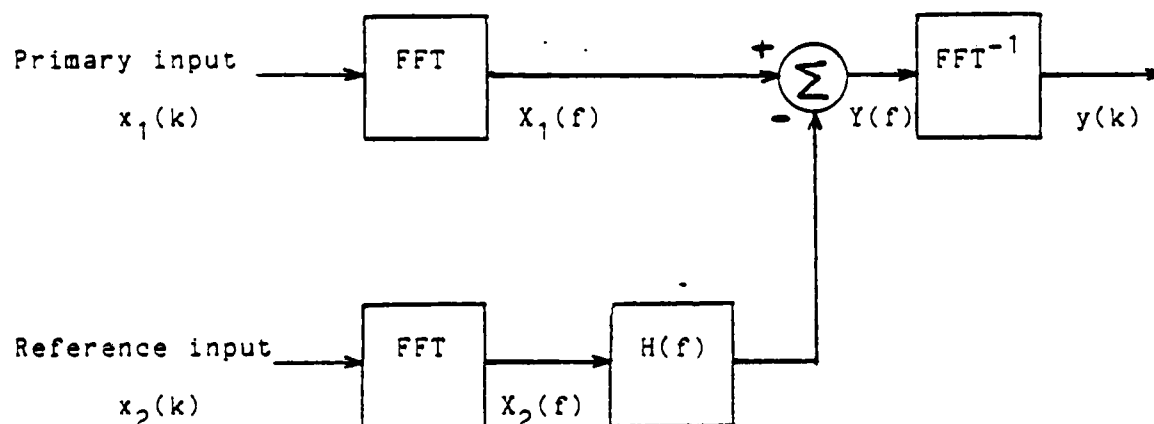


Figure E1

The Fast Fourier Transforms (FFT's) of the two signal inputs are calculated, the complex coefficients of the reference input FFT are multiplied by complex filter coefficients, $H(f)$. The

Enclosure 3 to Appendix D

.TITLE REALTI
.GLOBL REALTI

B. B. PETERSON, 9 MAY 84
IMPLEMENTS NOTCH FILTER IN REAL TIME.
CALLED FROM FORTRAN BY "CALL REALTI (NC,NW)"
NC = 256* CH. NO. + 16
NW = NO. WEIGHTS IN FILTER*2

IOA = 170404	:ADD OF CH. A D/A
IOB = 170406	:ADD OF CH. B D/A
IOC = 170400	:ADD OF CONTROL & STATUS REG.
IOI = 170402	:ADD OF A/D OUTPUT
REALTI: TST (R5)+	
MOV @(R5)+, @#IOC	:SETS CH NO AND EXT TRIG
MOV @(R5)+, R0	:R0 = NO OF WEIGHTS
MOV R0, R2	
LOOP0: MOV #0, DATA(R2)	:SET WEIGHT VECTOR TO ZERO
SCB R2, LOOP0	
LOOP1: MOV R0, R2	:INIT COUNTER
LOOP2: BIT #200, @#IOC	:TEST DONE BIT
SEQ LOOP2	
MOV @#IOI, R3	:READ A/D
MOV R2, @#IOA	:OUTPUT RAMP TO CH. A D/A TO CHECK SYNCH
SUB DATA(R2), R3	:R3=R3-WEIGHT(R2)
MOV R3, @#IOB	:FILTER OUTPUT TO CH. B D/A
ASR R3	:SHIFT R3 RIGHT 2 TIMES
ASR R3	
ADD R3, DATA(R2)	:ADJUST WEIGHT
DEC R2	
SCB R2, LOOP2	:DO THIS FOR NUMBER OF WEIGHTS
BR LOOP1	:GO BACK AND START AT BEGINNING OF VECTOR
LOOP4: RTS PC	
ATA: .BLKW 1000	:RESERVE ROOM FOR WEIGHTS
.END	

Enclosure 2 to Appendix D

```

STORE PRESENT POINTER IN DATA MEM 5
LP  SACL 5
READ IN OLD WEIGHT
    TRLR 4
READ A/D, PUT IN DATA MEM 6
    IN 8.2
THEN INTO ACC
    LAC 6.0
CONVERT FROM OFFSET BINARY TO 2'S COMPLEMENT
    XOR 7
SUBTRACT WEIGHT
    SUB 4.0
STORE RESULT IN DATA MEM 6
    SACL 6
CONVERT BACK TO OFFSET BINARY
    XOR 7
    SACL 8
OUTPUT TO D/A
    OUT 8.2
LOAD HIGH ACC WITH .25 OF OUTPUT
    LAC 6.14
NEW WEIGHT = OLD WEIGHT + .25 OUTPUT
    ADD 4
    SACL 4
LOAD ACC WITH LOCATION OF WEIGHT IN EXT MEM
    LAC 5.0
STORE NEW WEIGHT IN EXT MEM
    TRLW 4
INCREMENT ACC TO POINT TO NEXT WEIGHT
    ADDS 1
PUT AUX REG 0 OUT TO PORT 3 TO CHECK SYNCHRONIZATION
    SAR 0.9
    OUT 9.3
CHECK IF AT END OF WEIGHT VECTOR
    BANC SINT
IF SO, START AT BEGINNING
    LAR 0.3
    LAC 2.0
INT LARF 1
    INSERT DELAY TO PREVENT DOUBLE TRIGGER
    LARN 1.150
EL  NOP
    BANC DEL
    LARF 0
ENABLE INTERRUPT
    EINT
GO BACK AND WAIT FOR ANOTHER
    RET
END

```

Enclosure 2 to Appendix D

```
* NOTCH FILTER PROGRAM
* B. B. PETERSON. 19 JUN 84
* TRIGGER BY EXTERNAL INTERRUPT
* # OF WEIGHTS-1 MUST BE PUT IN DATA MEM 3
    B    INIT
    NOP
* INTERRUPT SERVICE
    B    ILP
* INITIALIZE
INIT LACK 0
* PUT 0 IN DATA MEM 0 TO USE ZEROING TABLE BELOW
    SACL 0
* SET UP A/D AND D/A CONTROL REGISTER
    LACK 7
    SACL 10
    OUT 10.0
* PUT 1000000000000000 IN DATA MEM 7 FOR OFFSET BINARY TO 2'S COMPLEMENT CON
    LACK 1
    SACL 1
    LAC 1.15
    SACL 7
* PUT 2 IN DATA MEM 1, INCREMENT IN TABLE
    LACK 2
    SACL 1
* PUT LOCATION OF START OF TABLE IN DATA MEM 2
    LACK 251
    SACL 2
* SET OVERFLOW MODE
    SQUM
    LARF 0
* ZERO TABLE
* PUT STARTING ADD OF TABLE IN ACCUMULATOR
    LAC 2.0
* PUT # OF WEIGHTS-1 IN AUX REG 0
    LAR 0.3
* PUT 0 IN THE EXT MEM POINTED TO BY THE ACC.
ZTR TBLW 0
* INCREMENT THE ACC BY 2
    ADDS 1
* DEC AUX REG 0 AND LOOP AGAIN IF NOT ZERO
    Banz ZTR
* MAIN LOOP TO START AT BEGINNING OF WEIGHT VECTOR
MLP LAR 0.3
    LAC 2.0
* ENABLE INTERRUPT
    EINT
* WAIT FOR INTERRUPT
WAIT NOP
    B    WAIT
```

Enclosure 1 to Appendix D

```

DO 1085 J = 1,NW
      J1=J+NW
      J2=J+2*NW
      J3=J+3*NW
      J4=J+4*NW
      J5=J+5*NW
      J6=J+6*NW
      IF (IFLAG.EQ.1) GO TO 1080
      WRITE (6,1060) J,IX(J),IX(J1),IX(J2),IX(J3),IX(J4),IX(J5),IX(J6)
      GO TO 1085
1080  WRITE (7,1060) J,IX(J),IX(J1),IX(J2),IX(J3),IX(J4),IX(J5),IX(J6)
1085  CONTINUE
1060  FORMAT (8I7)
C
C      CALCULATE AND PRINT OUT SIGNAL TO NOISE RATIO
C
1090  SNR=SGIN/SOUT
      WRITE(7,1100) SGIN,SOUT,SNP
1100  FORMAT (3F10.2)
      WRITE (7,1200)
1200  FORMAT (' NEW DATA? 1=YES,0=NO')
      READ (5,140) IFLAG2
      GO TO 50
1300  STOP
      END

```

Assembly Language Sampling subroutine called from main program

```

      .GLOBL XTSAMP
;
;      B. B. PETERSON, 17 NOV 83
;      SUBROUTINE TO TAKE A GIVEN NUMBER OF SAMPLES FROM THE DT2785
;      ANALOG I/O BOARD USING EXTERNAL TRIGGERING AT A MAXIMUM RATE OF
;      25,000 SAMP/S. THE SUBROUTINE IS CALLED BY "CALL XTSAMP(N,NC,IX)"
;      WHERE N= # OF SAMPLES,
;      NC = 256*CHANNEL # +16, AND IX IS THE DATA VECTOR.
;
      IOC=170400          ;LOCATION OF CONTROL REGISTER
      IOI=170402          ;LOCATION OF A/D OUTPUT REGISTER
XTSAMP: TST (R5)+
      MOV @ (R5)+,R0      ;R0=# OF SAMPLES
      MOV @ (R5)+,@#IOC   ;SET CH# AND EXTERNAL TRIGGER ENABLE
      MOV (R5)+,R3        ;R3=STARTING ADDRESS OF DATA VECTOR
LOOP1:  BIT #200,@#IOC     ;TEST DONE BIT
      BEQ LOOP1           ;LOOP IF NOT SET
      MOV @#IOI,R1        ;READ TO CLR DONE BIT,THROW AWAY 1ST SAMPLE
CONV:   BIT #200,@#IOC     ;TEST DONE BIT
      BEQ CONV           ;LOOP IF NOT SET
      MOV @#IOI,(R3)+     ;READ A/D AND PUT DATA IN MEMORY
      DEC R0
      BNE CONV           ;CHECK TO SEE IF N SAMPLES HAVE BEEN TAKEN
      RTS PC
      .END

```

Enclosure 1 to Appendix D

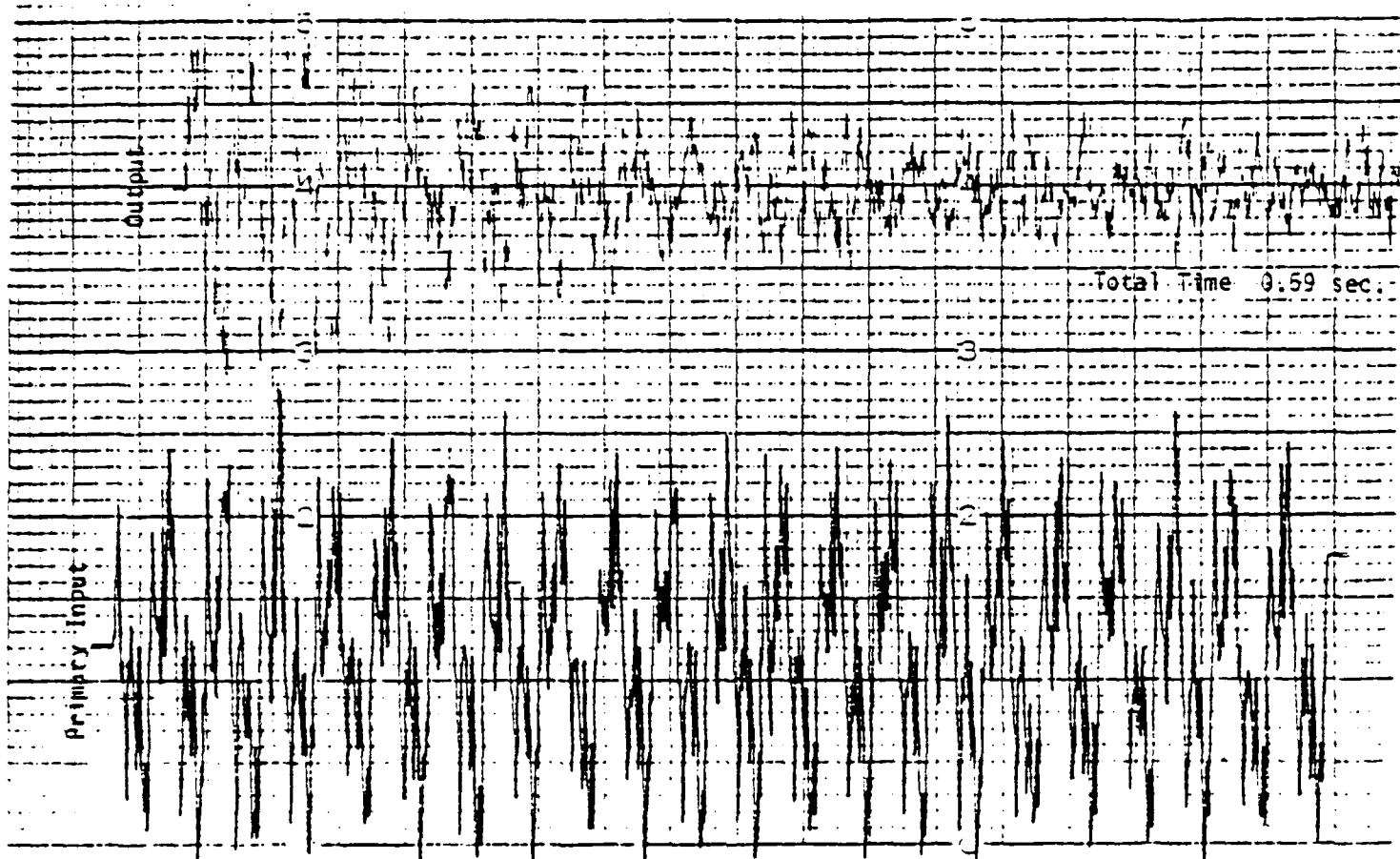
```

250          CONTINUE
C
C          GET N SAMPLES FROM SELECTED CHANNEL AND RETURN AS IX
C
300          CALL XTSAMP(N,NL,IX)
          WRITE (7,400)
400          FORMAT (' SAMPLING FINISHED')
C
C          INITIALIZE PLOTTER
C
          CALL GON
          CALL CLEAR
          CALL PLIMITS (.30,9.5,.3,6.5)
          CALL LIMITS(-FLOAT(N)/8.0,FLOAT(N),-1000.0,3000.0)
500          NP=(N/NW)
          CALL MOVE (-FLOAT(N)/9.0,50.0)
          CALL LABEL (' INPUT',2,0,0)
C
C          PLOT INPUT
C
          DO 525 I=1,N,ND
              CALL LINE (FLOAT(I),FLOAT(IX(I)))
525          CONTINUE
          CALL MOVE (-FLOAT(N)/9.0,2050.0)
          CALL LABEL (' OUTPUT',2,0,0)
C
C          CALCULATE AND PLOT OUTPUT
C
          DO 800 I=1,NP
              DO 700 J=1,NW
                  NS=(I-1)*NW+J
                  X=FLOAT(IX(NS))
                  Y=X-W(J)
                  W(J)=W(J)+Y*G
C
C          PLOT EVERY ND/TH. POINT
C
                  IF (ND1.LT.ND) GO TO 575
                  CALL LINE (FLOAT(NS),Y+2000.)
                  ND1=0
575          ND1=ND1+1
C
C          CALCULATE INPUT AND OUTPUT POWER FOR LAST TWO PERIODS
C
                  IF (I.LT.NP-2) GO TO 600
                  SQIN=SQIN+.5*X*X/NW
                  SOUT=SOUT+.5*Y*Y/NW
700          CONTINUE
800          CONTINUE
          CALL GOFF
          IF (IFLAG.EQ.0) GO TO 1090
C
C          PRINT OUT 7 PERIODS OF INPUT DATA IN 7 COLUMNS
C

```

Enclosure 1 to Appendix D

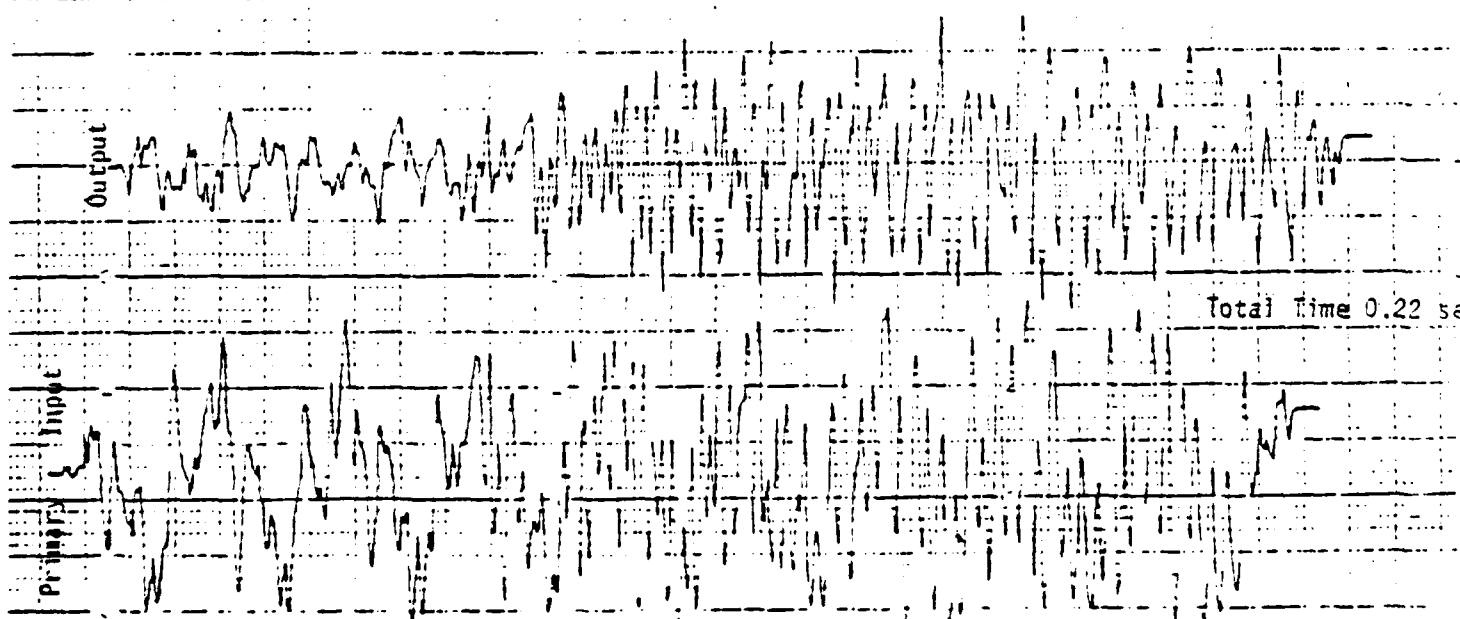
```
C      PROGRAM PERDIC
C      B. B. PETERSON
C      13 MAY 1984
C      PROGRAM TAKES GIVEN NUMBER OF SAMPLES OF AUDIO WAVEFORM AND
C      CALCULATES HOW MUCH OF WAVEFORM CAN BE CANCELLED BY PERIODIC
C      WAVEFORM AT FUNDAMENTAL FREQUENCY OF ENGINE
C
C      ARRAYS:
C          IX      SIGNAL INPUT
C          C      COMMUNICATIONS VECTOR FOR PLOTTING
C          W      WEIGHT VECTOR
C
C      INTEGER IX(10240)
C      COMMON C(20)
C      DIMENSION W(512)
C
C      SET FLAG TO GET DATA FIRST TIME
C
C      IFLAG2 = 1
C
C      ENTER PARAMETERS
C
C      50  WRITE (7,100)
C      100  FORMAT (' ENTER # SAMP, # SAMP/PLT, # WGHTS, CH NO, ADAP GAIN')
C      READ (5,200) N,ND,NW,NC,G
C
C      SELECT PRINTER AND PLOTTER OPTIONS
C
C      WRITE (7,125)
C      125  FORMAT (' OPTIONS:0=NO PRINTOUT,1=CRT,2=PRINTER')
C      READ (5,140) IFLAG
C      140  FORMAT (I3)
C      WRITE (7,145)
C      145  FORMAT (' 0=PLOTS ON PLOTTER,1=CRT')
C      READ (5,140) ID
C      C(10)=FLOAT(ID)
C
C      ZERO WEIGHT VECTOR AND INPUT AND OUTPUT POWER
C
C      DO 150 J=1,NW
C      150      W(J)=0.0
C      SGIN=0.0
C      SOUT=0.0
C      IF (N.LT.0) GO TO 1300
C      200  FORMAT (4I6,F8.5)
C
C      SKIP SAMPLING TO USE OLD DATA
C
C      IF (IFLAG2.EQ.0) GO TO 500
C      NC=256*NC+16
C
C      DELAY BEFORE SAMPLING
C
C      DO 250 J=1,30000
```



No Audio Signal in
Primary Input

Figure D7

Omniscene



Audio Signal
in Primary input

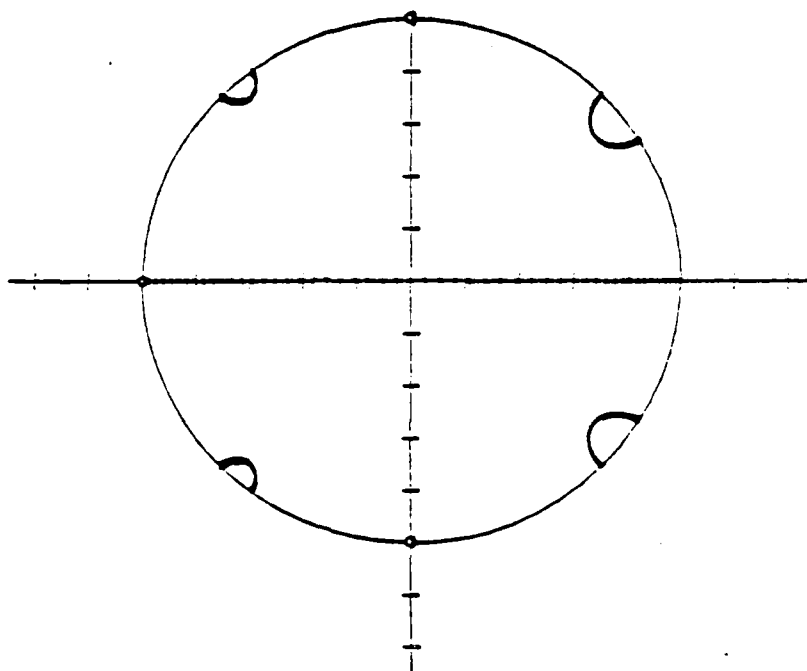
Figure D8

recorder. The plots in Figures D7 and D8 were produced using this program. Enclosure 1 is a listing of the program.

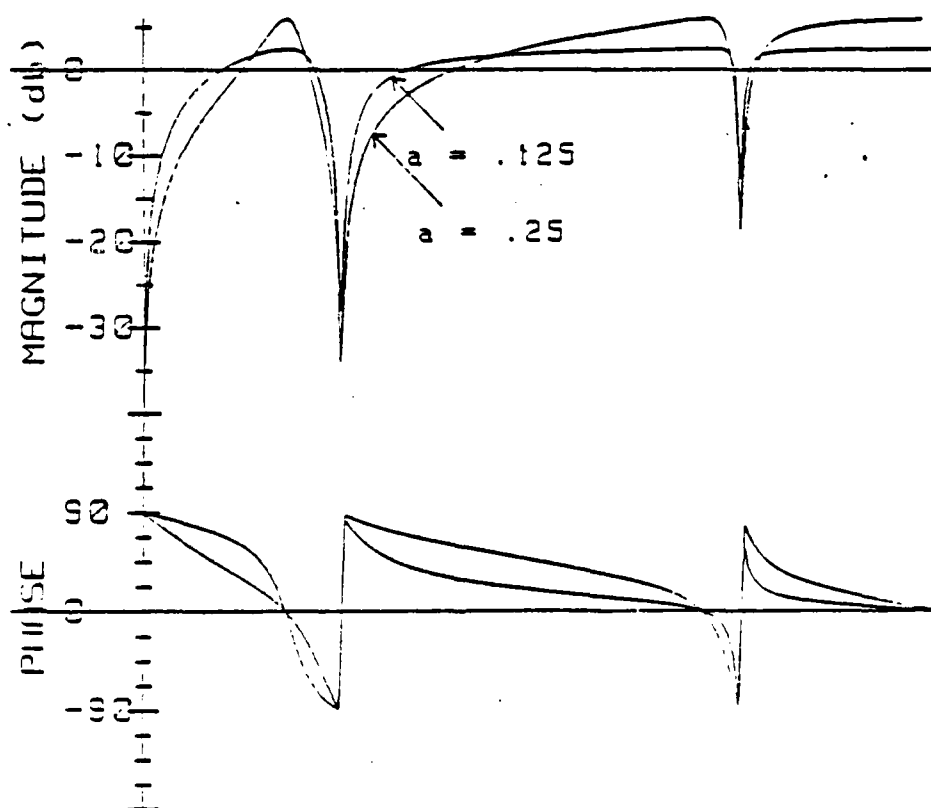
The filter was also implemented in real time on both the TMS 320 and the LSI-11/2. These assembly language program listings are enclosures 2 and 3 respectively. Because the TMS 320 has only 144 words of on-chip data memory, it was necessary to access external memory using the Table Read and Table Write instructions. These instructions require the address to be accessed to be placed in the low order accumulator. This results in relatively complicated code to implement a very simple algorithm. Despite this inefficient use of the TMS 320, the algorithm can still run at sampling rates of up to 150 kHz.

The MACRO-11 (PDP-11 assembly language) program in enclosure 3 illustrates the simplicity of the algorithm. On an LSI-11/2 this program will run at sampling rates of up to 11 kHz.

Cadet 1/c FAVERO is presently attempting to implement the filter on an AIM-65 microcomputer at sampling rates of at least 8 kHz.



ROOT LOCUS OF 8TH ORDER FILTER WHEN
REFERENCE INPUT IS A SQUARE WAVE
Figure D5



TRANSFER FUNCTION OF 8TH ORDER FILTER
WHEN REFERENCE INPUT IS A SQUARE WAVE
Figure D6

samples, the transfer function from $x_1(k)$ to $y(k)$ is given by:

$$\frac{Y(z)}{X_1(z)} = \frac{z^N - 1}{z^N + \sum_{j=0}^{N-1} a d_j z^j - 1}$$

where

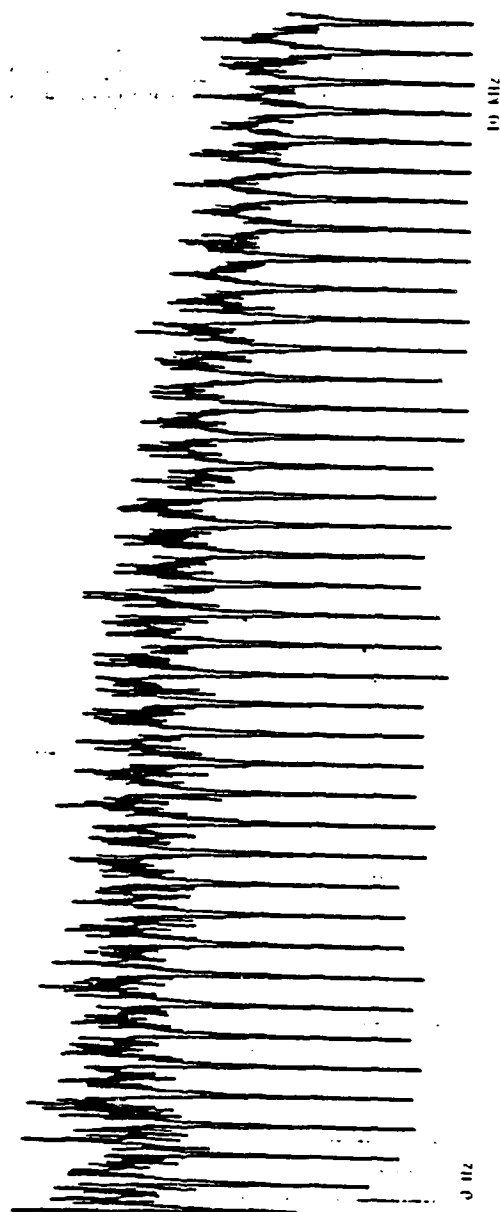
$$d_j = \sum_{i=0}^{N-1} x_2(i) x_2(i+j)$$

= N x the autocorrelation of x_2 .

Figure D5 is a root locus plot as the adaptive gain, a , is varied for $N = 8$ and for $x_2(k)$ a square wave and Figure D6 is the frequency response of the same filter. Now the width of each notch is a function the amplitude of the corresponding harmonic in $x_2(k)$ in addition to the adaptive gain. This suggests the possibility of using apriori knowledge of which harmonics are strongest in the noise component of $x_1(k)$ to specify an $x_2(k)$ for more effective filtering.

Another possible interesting application of this type of filter is for very narrow notch and bandpass recursive digital filters. The major advantage is that even with low precision fixed point arithmetic (8 bit, for example) and for poles virtually on the unit circle, the filter is completely stable.

The filter has been implemented in three different ways. First, using off-line processing in FORTRAN, a data vector is obtained by externally triggering the LSI-11/2 A/D converter and then the filter implemented and the data plotted on a strip chart



Frequency Response of Notch Filter
for $f_0 = 250$, $a = 0.25$ and $f_s = 40$ kHz

Figure 1A

Enclosure 1 to Appendix E

```

C      INITIALIZE COEFFICIENTS TO CALCULATE WEIGHT VECTOR
C      AND ZERO WEIGHT VECTOR
C
DO 40 I=1,NW2
    P2(I)=1.0E-10
    CR(I)=1.0E-10
    CI(I)=1.0E-10
    XH(I)=0.0
    YH(I)=0.0
40  CONTINUE
50  NBLOCS=N/NW
DO 140 J=1,NBLOCS
    JNW=(J-1)*NW

C      FILL VECTORS TO CALCULATE FFT

    PI=0.0
    PQ=0.0
    DO 120 I=1,NW2
        IA1= 2*I+JNW
        IA=IA1-1
        X1(I) = FLOAT(I1(IA)/205)
        Y1(I) = FLOAT(I1(IA1)/205)
        X2(I) = FLOAT(I2(IA)/205)
        Y2(I) = FLOAT(I2(IA1)/205)
120  CONTINUE
    CALL RFFT(NW,MW,NMAX,X1,Y1,WR,WI,1.0)
    CALL RFFT(NW,MW,NMAX,X2,Y2,WR,WI,1.0)
    DO 130 I=1,NW2

C      CALCULATE FFT OF OUTPUT

        XO(I)=X1(I)-YH(I)*X2(I)+YH(I)*Y2(I)
        YO(I)=Y1(I)-YH(I)*X2(I)-XH(I)*Y2(I)

C      CHECK FOR SMALL VALUES AND SET TO ZERO
C      TO PREVENT UNDERFLOWS

        IF (ABS(X1(I)).LT.1.0E-10) X1(I)=0.0
        IF (ABS(Y1(I)).LT.1.0E-10) Y1(I)=0.0
        IF (ABS(Y2(I)).LT.1.0E-10) Y2(I)=0.0
        IF (ABS(X2(I)).LT.1.0E-10) X2(I)=0.0

C      CALCULATE INPUT AND OUTPUT POWER

        PI=PI+X1(I)*X1(I)+Y1(I)*Y1(I)
        PQ=PQ+XO(I)*XO(I)+YO(I)*YO(I)

C      UPDATE WEIGHTS

        P2(I)=OMG*P2(I)+G*(X2(I)*X2(I)+Y2(I)*Y2(I))
        CR(I)=OMG*CR(I)+G*(X1(I)*X2(I)+Y1(I)*Y2(I))
        CI(I)=OMG*CI(I)+G*(X2(I)*Y1(I)-X1(I)*Y2(I))
        XH(I)=CR(I)/P2(I)

```

Enclosure 1 to Appendix E

```

                                YH(I)=CI(I)/P2(I)
130      CONTINUE
C
C      CALCULATE OUTPUT
C
C      CALL REFT(NN,MN,NMAX,XO,YO,WR,WI,-1.0)
C
C      PQ=PQ/NN
C      PI=PI/NW
C      WRITE (7,137) J,PI,PQ
137      FORMAT (16,2F12.2)
140      CONTINUE
      GO TO 5
      END
```

Enclosure 1 to Appendix E FFT subroutines called from main program

```

C      FFT SUBROUTINES
C      R. B. PETERSON, 5 FEB 85
C
C      SUBROUTINE RFFT(N,M,NMAX,X,Y,WR,WI,DI)
C
C      SUBROUTINE TO CALCULATE N POINT REAL FFT USING
C      N/2 POINT COMPLEX FFT
C
C      REF: "TMS 320 DIGITAL SIGNAL PROCESSING SEMINAR NOTES,"
C      TEXAS INSTRUMENTS, INC., 1983.
C
C      N= # OF REAL DATA POINTS = 2*M
C      X IS REAL PART OF FFT (N/2 VECTOR) Y IS IMAG PART
C      WR AND WI ARE VECTORS OF COSINES AND SINES OF SIZE NMAX
C      DI= 1. FOR FORWARD FFT, -1. FOR INVERSE.
C      FOR FORWARD FFT X AND Y SHOULD BE GENERATED BY
C      X(I) = DATA(2*I-1) AND Y(I) = DATA(2*I)
C      WHERE DATA() IS THE INPUT VECTOR OF REAL POINTS
C
C      REAL X(1),Y(1),WR(1),WI(1)
C      N2=N/2
C      MM1 = M-1
C      IF (DI.LT.0.0) GO TO 10
C      CALL FFT2(N2,MM1,NMAX,X,Y,WR,WI,1.0).
10     NMON=NMAX/N
C      N4=N2/2
C      Y(N2+1)=Y(1)
C      X(N2+1)=X(1)
C      IP=1
C      DO 20 I= 1, N4+1
C         NMI=N2-I+2
C         RA=X(I)+X(NMI)
C         XA=Y(I)-Y(NMI)
C         RB=X(NMI)-X(I)
C         YB=(-Y(I)-Y(NMI))
C         CPH=WR(IP)
C         SPH=-WI(IP)
C         IF (DI.GT.0.0) GO TO 15
C         RB=-RB
C         YB=-YB
C         SPH=-SPH
15     IP=IP+NMON
C         RC=SPH*RB-CPH*YB
C         XC=SPH*YB+CPH*RB
C         X(I)=.5*(RA+RC)
C         Y(I)=.5*(XA+XC)
C         X(NMI)=.5*(RA-RC)
C         Y(NMI)=.5*(XC-XA)
20     CONTINUE
C      IF (DI.GT.0.0) GO TO 1000
C      CALL FFT2(N2,MM1,NMAX,X,Y,WR,WI,-1.0)
1000    RETURN
C      END

```

Enclosure 1 to Appendix E FFT subroutines called from main program

```

C      COMPLEX FFT WITH TABLE LOOKUP
C      REF: C. S. BURRIS AND T. W. PARKS, "DFT/FFT AND CONVOLUTION
C      ALGORITHMS," JOHN WILEY & SONS, NEW YORK, 1984
C
C      CALLING VARIABLES:
C          N          NUMBER OF COMPLEX POINTS
C          M          LOG2 OF N
C          NMAX       MAXIMUM VALUE OF N FOR INTERPRETING LOOKUP TABLE
C          X & Y      REAL AND IMAGINARY PARTS OF FFT
C          WR         TABLE OF COSINES
C          WI         TABLES OF -SINES
C          DI         +1.0 FOR FORWARD FFT, -1.0 FOR INVERSE
C
C      SUBROUTINE FFT2(N,M,NMAX,X,Y,WR,WI,DI)
C      REAL Y(1),X(1),WR(1),WI(1)
C      NMON=NMAX/N
C      N2=N
C      DO 100 K=1,M
C          N1=N2
C          N2=N2/2
C          IE=N/N1
C          IA=1
C          DO 50 J=1,N2
C              IA1=(IA-1)*NMON+1
C              C=WR(IA1)
C              S=-DI*WI(IA1)
C              IA=IA+IE
C              DO 40 I=J,N,N1
C                  L=I+N2
C                  XT=X(I)-X(L)
C                  X(I)=X(I)+X(L)
C                  YT=Y(I)-Y(L)
C                  Y(I)=Y(I)+Y(L)
C                  X(L)=C*XT+S*YT
C                  Y(L)=C*YT-S*XT
C          40      CONTINUE
C      50      CONTINUE
C      100     CONTINUE
C      J=1
C      N1=N-1
C      DO 104 I=1,N1
C          IF (I.GE.J) GO TO 101
C          XT=X(J)
C          X(J)=X(I)
C          X(I)=XT
C          YT=Y(J)
C          Y(J)=Y(I)
C          Y(I)=YT
C      101     K=N/2
C      102     IF (K.GE.J) GO TO 103
C              J=J-K
C              K=K/2
C              GO TO 102
C      103     J=J+K

```


Enclosure 1. to Appendix E FFT subroutines called from main program

```
104      CONTINUE
        IF (DI.GT.0.0) GO TO 106
        DO 105 I=1,N
          X(I)=X(I)/N
          Y(I)=Y(I)/N
105      CONTINUE
106      RETURN
        END

C
SUBROUTINE PACK (N,NMAX,IX,X,Y,IW,WIN)
C
C      PUTS INTEGER DATA IN REAL ARRAYS SUITABLE FOR FFT SUB ABOVE
C      MULTIPLIES EACH DATA POINT BY HAMMING WINDOW
C

      REAL X(1),Y(1),WIN(1)
      INTEGER IX(1)
      NMON=NMAX/N
      N2=N/2
      DO 10 I=1,N2
        I2=I*2
        X(I)=FLOAT (IX(I2-1))
        Y(I)=FLOAT (IX(I2))
        IF (IW.EQ.0) GO TO 10
        I3=(I2-2)*NMON+1
        X(I)=X(I)*WIN(I3)
        Y(I)=Y(I)*WIN(NMON+I3)
10      CONTINUE
      RETURN
      END
```

Appendix F LMS Adaptive Filter FORTRAN Program where Audio Signal is added
in Software

```

PROGRAM ADAPT
C      B. B. PETERSON
C      PROGRAM READS IN ANALOG DATA FROM AN FM TUNER
C      DECK USING SUBROUTINE XTSMF2. IT THEN SAMPLES TWO CHANNELS FROM
C      A TAPE DECK. THE FM TUNER SIGNAL IS MULT. BY A GAIN AND ADDED
C      TO ONE NOISE CHANNEL TO GENERATE THE PRIMARY INPUT. THE OTHER NOISE
C      CHANNEL IS ADAPTIVELY FILTERED AND SUBTRACTED FROM THE DELAYED
C      PRIMARY INPUT. THE WEIGHTS IN THE ADAPTIVE FILTER ARE ADJUSTED SO
C      AS TO MINIMIZE THE OUTPUT POWER. THE REFERENCE INPUT, THE PRIMARY
C      INPUT, THE FM TUNER SIGNAL AND THE OUTPUT ARE PLOTTED ON THE CRT,
C      THE HP 7470 PLOTTER OR BOTH. THE SIGNAL TO NOISE RATIOS OF BOTH
C      INPUT AND OUTPUT ARE CALCULATED AFTER THE WEIGHTS HAVE HAD SOME
C      TIME TO CONVERGE.
C
C      ARRAYS:
C          IY      PRIMARY INPUT
C          IX      REFERENCE INPUT
C          ISIG     INTEGER SIGNAL VECTOR
C          H        WEIGHT VECTOR
C          C        COMMUNICATIONS VECTOR FOR PLTLIB
C
C      INTEGER IX(4000), IY(4000), ISIG(4000)
C      REAL H(100)
C      COMMON C(20)
C
C      INPUT PARAMETERS
C
C      50      WRITE (7,100)
C      100     FORMAT (' NEW DATA?, RESET WEIGHT VECTOR? 1=YES,0=NO')
C      READ (5,200) IF1,IF2
C      200     FORMAT (2I6)
C      IF (IF1.EQ.0 .AND. IF2.EQ.0) GO TO 450
C      WRITE (7,300)
C      300     FORMAT (' # SAMP,REF INP,PRI INP,SIG INP,# WTS')
C      READ (5,325) N,NC1,NC2,NC3,NW
C      325     FORMAT (5I6)
C      NT=NW/2
C      N2=N/2
C      WRITE (7,350)
C      350     FORMAT (' PLOT RAW DATA?,0=7470,1=GIGI,2=BOTH,-1=NO PLOT,STEP')
C      READ (5,200) IC10,ISTP
C      STP=FLOAT(ISTP)
C      IF (IC10.LT.0) GO TO 370
C      C(10)=FLOAT (IC10)
C      370     NC1=256*NC1+16
C      NC2=256*NC2+16
C      NC3=256*NC3+16
C      450     WRITE (7,500)
C      500     FORMAT (' ADAPTIVE GAIN (E FORMAT),SIGNAL GAIN (I FORMAT)')
C      READ (5,600) G,IG
C      600     FORMAT (E14.4,I6)
C      IF (IF2.EQ.0) GO TO 800
C
C      ZERO WEIGHT VECTOR

```

```

C      DO 700 I=1,NW
700      H(I)=0.0
800      IF (IF1.EQ.0) GO TO 825
C
C      DELAY BEFORE SAMPLING
C
C      DO 850 I=1,10000
850      CONTINUE
C
C      SAMPLE SIGNAL FROM TUNER
C
C      CALL XTSMPT(N,NC3,NC2,ISIG,IY)
C
C      SAMPLE TWO CHANNELS FROM TAPE DECK
C
C      CALL XTSMPT(N,NC1,NC2,IX,IY)
825      IF (IC10.LT.0) GO TO 900
      CALL PLIMITS (0.0,10.0,0.0,7.0)
      CALL LIMITS (0.0,FLOAT(N),-.8*STP,3.6*STP)
C      IF (IF1.EQ.0) GO TO 900
      DO 875 I=1,N
          CALL LINE (FLOAT(I),FLOAT(IX(I)))
875      CONTINUE
      CALL MOVE (1.0,STP)
      DO 890 I=1,N
          ISIG(I)=IG+ISIG(I)
          IY(I)=IY(I)+ISIG(I)
          CALL LINE (FLOAT(I),FLOAT(ISTP+IY(I)))
890      CONTINUE
      CALL MOVE (1.0,2*STP)
      OFFSET= 2.0*STP
      DO 895 I=1,N
          CALL LINE (FLOAT(I),FLOAT(ISIG(I))+OFFSET)
895      CONTINUE
      OFFSET=3.0*STP
      CALL MOVE (1.0,OFFSET)
C
C      ZERO SQUARE      PUT NOISE,OUTPUT NOISE, AND DESIRED OUTPUT.
C
C      900      SGIN=0.0
      SGOUT=0.0
      SGSIG=0.0
C
C      MAIN ADAPTIVE FILTER LOOP
C
C      DO 1000 I=NW+1,N
          FOUT=0.0
C
C      CALCULATE FILTER OUTPUT
C
C      DO 950 J=1,NW
          IS=(I-J)
950      FOUT=FOUT+H(J)*IX(IS)

```

```

ID=(I-NT)
OUT= IY(ID)-FOUT

```

```

SQUARE AND LOW PASS FILTER INPUT AND OUTPUT
FOR LAST HALF OF DATA

```

```

IF (I.LT.N2) GO TO 980
SQOUT=.98*SQOUT+.02*(OUT-FLOAT(ISIG(ID)))*.2
RIX=FLOAT(IY(ID)-ISIG(ID))
SQIN=.98*SQIN+.02*RIX*RIX
SQSIG=.98*SQSIG+.02*FLOAT(ISIG(I))*FLOAT(ISIG(I))
SNRIN=SQSIG/SQIN
SNROUT=SQSIG/SQOUT

```

```

PLOT OUTPUT

```

```

CALL LINE (FLOAT(I),OUT+OFFSET)

```

```

ADAPT WEIGHT VECTOR

```

```

DO 975 J=1,NW
    IS=(I-J)
    H(J)=H(J)+G*OUT*IX(IS)

```

```

CONTINUE
CALL TURNOF
WRITE (7,1050) SNRIN,SNROUT
FORMAT (2F10.5)
GO TO 50
STOP
END

```

References

- [1] B. Widrow, et. al. "Adaptive Noise Cancelling: Principles and Applications," *Proceedings of the IEEE*, Vol. 63, No. 12, pp. 1692-1716, December 1975.

- [2] J. R. Glover, Jr., "Adaptive Noise Cancelling Applied to Sinusoidal Interferences," *IEEE Trans. on Acoustics, Speech, and Signal Processing*, Vol. ASSP-25, No. 6, pp. 484-491, December 1977.

- [3] R. H. Frazier, et. al., "Enhancement of Speech by Adaptive Filtering," *Proceedings of the IEEE International Conference on Acoustics, Speech, and Signal Processing*, pp. 251-253, April 1976.

- [4] J. S. Lim, A. V. Oppenheim, and L. D. Braida, "Evaluation of an Adaptive Comb Filtering Method for Enhancing Speech Degraded by White Noise Addition," *IEEE Trans. on Acoustics, Speech, and Signal Processing*, Vol. ASSP-26, No. 5, pp. 354-358, August 1978.

- [5] M. Dentino, J. McCool, and B. Widrow, "Adaptive Filtering in the Frequency Domain," *Proc. IEEE*, Vol. 66, No. 12, pp. 1658-1659, December 1978.

- [6] B. Widrow, J. McCool, and M. Ball, "The Complex LMS Algorithm," *Proc. IEEE*, Vol. 63, No. 4, pp. 719-720, April 1975.

- [7] B. Gold and J. Tierney, "Vocoder Analysis Based on Properties of the Human Auditory System," Technical Report 670, Lincoln Laboratory, Massachusetts Institute of Technology, December 1983.
- [8] B. Gold, "Robust Vocoding: Is Pitch the Problem?" IEEE Digital Signal Processing Workshop, p. 4.1, Chatham, MA, October 8-10, 1984.
- [9] M. R. Sambur and N. S. Jayant, "LPC Analysis/Synthesis from Speech Inputs Containing Quantizing Noise or Additive White Noise," IEEE Trans. on Acoustics, Speech and Signal Processing, Vol. ASSP-24, No. 6, pp. 488-494, December 1976.
- [10] H. Kobatake, J. Inari, and S. Kakuta, "Linear Predictive Coding of Speech Signals in a High Ambient Noise Environment," IEEE International Conference on Acoustics, Speech, and Signal Processing, pp. 472-475, June 1978.
- [11] W. D. Voiers, A. D. Sharply, and C. J. Hemsoth, "Research on Diagnostic Evaluation of Speech Intelligibility," Final Report, Contract No. AF19628-70-C-0182, Air Force Cambridge Research Laboratories, 1973.
- [12] T. E. Tremain, "The Government Standard Linear Predictive Coding Algorithm: LPC-10," Speech Technology, pp. 40-49, April 1982.

[13] B. B. Peterson, "A Proposal for Research on Application of Adaptive Noise Cancellation to Coast Guard Voice Communications," U. S. Coast Guard Academy, April 1983.

[14] TMS 32020 Users Guide, Preliminary Draft, Texas Instruments, Inc., Houston, Texas, January 1985.

[15] S. W. Director, Circuit Theory: A Computational Approach, John Wiley & Sons, Inc, New York, 1975.

[16] B. B. Peterson, "Adaptive Noise Cancellation Applied to Coast Guard Voice Communications, Interim Report," U. S. Coast Guard Academy, July 1984.

[17] Details on Signal Processing, Issue 1, Texas Instruments, Inc. March 1984.

[18] Microcomputers and Memories, Digital Equipment Corporation, 1982.

[19] Model 562 Data Sheet, Shure Brothers, Inc. 1980.

[20] Robert Olson, "Noise Cancelling Boom Microphone," Electronics Systems Information Bulletin, Commandant (G-TES), U. S. Coast Guard, November 1983.

[21] J. J. Wolcin, "A General Method and FORTRAN program For the Design of Recursive Digital Filters," NUSC Technical Report 4629, Naval Underwater Systems Center, New London Laboratory, September 1973.

[22] V. R. Viswanathan, et. al. "Multisensor Speech Input," RADC-TR-83-274, Rome Air Development Center, December 1983.

[23] C. F. Teacher and A. J. Brouns, "Microphone for Very High Noise Environments," IEEE Digital Signal Processing Workshop, pp. 4.2.1-2, Chatham, MA, October 1984.

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